

CONVERGENCE OF A FINITE-VOLUME SCHEME FOR A DEGENERATE CROSS-DIFFUSION MODEL FOR ION TRANSPORT

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ABSTRACT. An implicit Euler finite-volume scheme for a degenerate cross-diffusion system describing the ion transport through biological membranes is analyzed. The strongly coupled equations for the ion concentrations include drift terms involving the electric potential, which is coupled to the concentrations through the Poisson equation. The cross-diffusion system possesses a formal gradient-flow structure revealing nonstandard degeneracies, which lead to considerable mathematical difficulties. The finite-volume scheme is based on two-point flux approximations with “double” upwind mobilities. It preserves the structure of the continuous model like nonnegativity, upper bounds, and entropy dissipation. The degeneracy is overcome by proving a new discrete Aubin-Lions lemma of “degenerate” type. Under suitable assumptions, the existence and uniqueness of bounded discrete solutions, a discrete entropy inequality, and the convergence of the scheme is proved. Numerical simulations of a calcium-selective ion channel in two space dimensions indicate that the numerical scheme is of first order.

1. INTRODUCTION

The ion transport through biological channels plays an important role in all living organisms. On a macroscopic level, the transport can be described by nonlinear partial differential equations for the ion concentrations (or, more precisely, volume fractions) and the surrounding electric potential. A classical model for ion transport are the Poisson-Nernst-Planck equations [25], which satisfy Fick’s law for the fluxes. However, this approach does not include size exclusion effects in narrow ion channels. Taking into account the finite size of the ions, one can derive from an on-lattice model in the diffusion limit another set of differential equations with fluxes depending on the gradients of all species [10, 27]. These nonlinear cross-diffusion terms are common in multicomponent systems [23, Chapter 4]. In this paper, we propose an implicit Euler finite-volume discretization of the resulting cross-diffusion system. The scheme is designed in such a way that the nonnegativity and

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upper bound of the concentrations as well as the entropy dissipation is preserved on the discrete level.

More specifically, the evolution of the concentrations u_i and fluxes \mathcal{F}_i of the i th ion species is governed by the equations

$$(1) \quad \partial_t u_i + \operatorname{div} \mathcal{F}_i = 0, \quad \mathcal{F}_i = -D_i(u_0 \nabla u_i - u_i \nabla u_0 + u_0 u_i \beta z_i \nabla \Phi) \quad \text{in } \Omega, \quad t > 0,$$

for $i = 1, \dots, n$, where $u_0 = 1 - \sum_{i=1}^n u_i$ is the concentration (volume fraction) of the electro-neutral solvent, $D_i > 0$ is a diffusion coefficient, $\beta > 0$ is the (scaled) inverse thermal voltage, and $z_i \in \mathbb{R}$ the charge of the i th species. Observe that we assumed Einstein's relation which says that the quotient of the diffusion and mobility coefficients is constant, and we call this constant $1/\beta$. The electric potential is determined by the Poisson equation

$$(2) \quad -\lambda^2 \Delta \Phi = \sum_{i=1}^n z_i u_i + f \quad \text{in } \Omega,$$

where λ^2 is the (scaled) permittivity constant and $f = f(x)$ is a permanent background charge density. Equations (1) and (2) are solved in a bounded domain $\Omega \subset \mathbb{R}^d$.

In order to match experimental conditions, the boundary $\partial\Omega$ is supposed to consist of two parts, the insulating part Γ_N , on which no-flux boundary conditions are prescribed, and the union Γ_D of boundary contacts with external reservoirs, on which the concentrations are fixed. The electric potential is prescribed at the electrodes on Γ_D . This leads to the mixed Dirichlet-Neumann boundary conditions

$$(3) \quad \mathcal{F}_i \cdot \nu = 0 \quad \text{on } \Gamma_N, \quad u_i = \bar{u}_i \quad \text{on } \Gamma_D, \quad i = 1, \dots, n,$$

$$(4) \quad \nabla \Phi \cdot \nu = 0 \quad \text{on } \Gamma_N, \quad \Phi = \bar{\Phi} \quad \text{on } \Gamma_D,$$

where the boundary data $(\bar{u}_i)_{1 \leq i \leq n}$ and $\bar{\Phi}$ can be defined on the whole domain Ω . Finally, we prescribe the initial conditions

$$(5) \quad u_i(\cdot, 0) = u_i^I \quad \text{in } \Omega, \quad i = 1, \dots, n.$$

The main mathematical difficulties of equations (1) are the strong coupling and the fact that the diffusion matrix $(A_{ij}(u))$, defined by $A_{ij}(u) = D_i u_i$ for $i \neq j$ and $A_{ii}(u) = D_i(u_0 + u_i)$ is not symmetric and not positive definite. It was shown in [10, 22] that system (1) possesses a formal gradient-flow structure. This means that there exists a (relative) entropy functional $H[u] = \int_{\Omega} h(u) dx$ with the entropy density

$$h(u) = \sum_{i=0}^n \int_{\bar{u}_i}^{u_i} \log \frac{s}{\bar{u}_i} ds + \frac{\beta \lambda^2}{2} |\nabla(\Phi - \bar{\Phi})|^2,$$

where $u = (u_1, \dots, u_n)$ and $u_0 = 1 - \sum_{i=1}^n u_i$, such that (1) can be formally written as

$$\partial_t u_i = \operatorname{div} \left(\sum_{j=1}^n B_{ij} \nabla w_j \right),$$

where $B_{ii} = D_i u_0 u_i$, $B_{ij} = 0$ for $i \neq j$ provide a diagonal positive definite matrix, and w_j are the entropy variables, defined by

$$\frac{\partial h}{\partial u_i} = w_i - \bar{w}_i, \quad \text{where}$$

$$w_i = \log \frac{u_i}{u_0} + \beta z_i \Phi, \quad \bar{w}_i = \log \frac{\bar{u}_i}{\bar{u}_0} + \beta z_i \bar{\Phi}, \quad i = 1, \dots, n.$$

We refer to [20, Lemma 7] for the computation of $\partial h / \partial u_i$.

The entropy structure of (1) is useful for two reasons. First, it leads to L^∞ bounds for the concentrations. Indeed, the transformation $(u, \Phi) \mapsto w$ to entropy variables can be inverted, giving $u = u(w, \Phi)$ with

$$u_i(w, \Phi) = \frac{\exp(w_i - \beta z_i \Phi)}{1 + \sum_{j=1}^n \exp(w_j - \beta z_j \Phi)}, \quad i = 1, \dots, n.$$

Then u_i is positive and bounded from above, i.e.

$$(6) \quad u \in \mathcal{D} = \left\{ u \in (0, 1)^n : \sum_{i=1}^n u_i < 1 \right\}.$$

This yields L^∞ bounds without the use of a maximum principle. Second, the entropy structure leads to gradient estimates via the entropy inequality

$$\frac{dH}{dt} + \frac{1}{2} \int_{\Omega} \sum_{i=1}^n D_i u_0 u_i |\nabla w_i|^2 dx \leq C,$$

where the constant $C > 0$ depends on the Dirichlet boundary data. Because of

$$(7) \quad \sum_{i=1}^n u_0 u_i \left| \nabla \log \frac{u_i}{u_0} \right|^2 = 4 \sum_{i=1}^n u_0 |\nabla u_i^{1/2}|^2 + 4 |\nabla u_0^{1/2}|^2 + |\nabla u_0|^2,$$

we achieve gradient estimates for $u_0^{1/2} u_i$ and $u_0^{1/2}$. Since u_0 may vanish locally, this does not give gradient bounds for u_i , which expresses the degenerate nature of the cross-diffusion system. As a consequence, the flux has to be formulated in the terms of gradients of $u_0^{1/2} u_i$ and $u_0^{1/2}$ only, namely

$$(8) \quad \mathcal{F}_i = -D_i (u_0^{1/2} \nabla (u_0^{1/2} u_i) - 3u_0^{1/2} u_i \nabla u_0^{1/2} + u_0 u_i \beta z_i \nabla \Phi).$$

The challenge is to derive a discrete version of this formulation. It turns out that (24) below is the right formulation in our context (assuming vanishing drift parts).

Our aim is to design a numerical approximation of (1) which preserves the structural properties of the continuous equations. This suggests to use the entropy variables as the unknowns, as it was done in our previous work [20] with simulations in one space dimension. Unfortunately, we have not been able to perform a numerical convergence analysis with these variables. The reason is that we need discrete chain rules in order to formulate (7) on the discrete level and these discrete chain rules seem to be not easily available. Therefore, we use the original variables u_i for the numerical discretization. Interestingly, we are still able to prove that the scheme preserves the nonnegativity, upper bound, and

entropy inequality. However, the upper bound comes at a price: We need to assume that all diffusion coefficients D_i are the same. Under this assumption, $u_0 = 1 - \sum_{i=1}^n u_i$ solves a drift-diffusion equation for which the (discrete) maximum principle can be applied. It is not surprising that the L^∞ bound can be shown only under an additional condition, since cross-diffusion systems usually do not allow for a maximum principle.

The key observation for the numerical discretization is that the fluxes can be written on each cell in a “double” drift-diffusion form, i.e., both $\mathcal{F}_i = -D_i(u_0 \nabla u_i - u_i V_i)$ and $V_i = \nabla u_0 - \beta z_i u_0 \nabla \Phi$ have the structure $\nabla v + vF$, where ∇v is the diffusion term and vF is the drift term. We discretize \mathcal{F} and V by using a two-point flux approximation with “double” upwind mobilities.

Our analytical results are stated and proved for no-flux boundary conditions on $\partial\Omega$. Mixed Dirichlet-Neumann boundary conditions could be prescribed as well, but the proofs would become even more technical. The main results are as follows.

- We prove the existence of solutions to the fully discrete numerical scheme (Theorem 1). If the drift part vanishes, the solution is unique. The existence proof uses a topological degree argument in finite space dimensions, while the uniqueness proof is based on the entropy method of Gajewski [18], recently extended to cross-diffusion systems [20, 28].
- Thanks to the “double” upwind structure, the scheme preserves the nonnegativity and upper bound for the concentrations (at least if $D_i = D$ for all i). Moreover, convexity arguments show that the discrete entropy is dissipated with a discrete entropy production analogous to (7) (Theorem 2). The proof of the discrete entropy inequality only works if the drift term vanishes, since we need to control a discrete version of the sum $\sum_{i=1}^n u_i$ from below; see the discussion after Theorem 2.
- The discrete solutions converge to the continuous solutions to (1) as the mesh size tends to zero (Theorem 3). The proof is based on a priori estimates obtained from the discrete entropy inequality. The compactness is derived from a new discrete Aubin-Lions lemma, which takes into account the nonstandard degeneracy of the equations; see Lemma 10 in the appendix.
- Numerical experiments for a calcium-selective ion channel in two space dimensions show the dynamical behavior of the solutions and their large-time asymptotics to the equilibrium. The tests indicate that the order of convergence in the L^1 norm is one.

In the literature, there exist some results on finite-volume schemes for cross-diffusion systems. An upwind two-point flux approximation similar to our discretization was recently used in [1] for a seawater intrusion cross-diffusion model. A two-point flux approximation with a nonlinear positivity-preserving approximation of the cross-diffusion coefficients, modeling the segregation of a two-species population, was suggested in [4], assuming positive definiteness of the diffusion matrix. The Laplacian structure of the population model (still for positive definite matrices) was exploited in [24] to design a convergent *linear* finite-volume scheme, which avoids fully implicit approximations. A semi-implicit finite-volume discretization for a biofilm model with a nonlocal time integrator was proposed in [26].

Finite-volume schemes for cross-diffusion systems with nonlocal (in space) terms were also analyzed; see, for instance, [3] for a food chain model and [2] for an epidemic model. Moreover, a finite-volume scheme for a Keller-Segel system with additional cross diffusion and discrete entropy dissipation property was investigated in [7]. All these models, however, do not include volume filling and do not possess the degenerate structure explained before.

The paper is organized as follows. The numerical scheme and the main results are presented in Section 2. In Section 3, the existence and uniqueness of bounded discrete solutions are shown. We prove the discrete entropy inequality and further a priori estimates in Section 4, while Section 5 is concerned with the convergence of the numerical scheme. Numerical experiments are given in Section 6 in order to illustrate the order of convergence and the long time behavior of the scheme. For the compactness arguments, we need two discrete Aubin-Lions lemmas which are proved in the appendix.

2. NUMERICAL SCHEME AND MAIN RESULTS

2.1. Notations and definitions. We summarize our general hypotheses on the data:

- (H1) Domain: $\Omega \subset \mathbb{R}^d$ ($d = 2$ or $d = 3$) is an open, bounded, polygonal domain with $\partial\Omega = \Gamma_D \cup \Gamma_N \in C^{0,1}$, $\Gamma_D \cap \Gamma_N = \emptyset$.
- (H2) Parameters: $T > 0$, $D_i > 0$, $\beta > 0$, and $z_i \in \mathbb{R}$, $i = 1, \dots, n$.
- (H3) Background charge: $f \in L^\infty(\Omega)$.
- (H4) Initial and boundary data: $u_i^I \in L^\infty(\Omega)$, $\bar{u}_i \in H^1(\Omega)$ satisfy $u_i^I \geq 0$, $\bar{u}_i \geq 0$ and $1 - \sum_{i=1}^n u_i^I \geq 0$, $1 - \sum_{i=1}^n \bar{u}_i \geq 0$ in Ω for $i = 1, \dots, n$, and $\bar{\Phi} \in H^1(\Omega) \cap L^\infty(\Omega)$.

For our main results, we need additional technical assumptions:

- (A1) $\partial\Omega = \Gamma_N$, i.e., we impose no-flux boundary conditions on the whole boundary.
- (A2) The diffusion constants are equal, $D_i = D > 0$ for $i = 1, \dots, n$.
- (A3) The drift terms are set to zero, $\Phi \equiv 0$.

Remark 1 (Discussion of the assumptions). Assumption (A1) is supposed for simplicity only. Mixed Dirichlet-Neumann boundary conditions can be included in the analysis (see, e.g., [20]), but the proofs become even more technical. Mixed boundary conditions are chosen in the numerical experiments; therefore, the numerical scheme is defined for that case. Assumption (A2) is needed for the derivation of an upper bound for the solvent concentration. Indeed, when $D_i = D$ for all i , summing (1) over $i = 1, \dots, n$ gives

$$\partial_t u_0 = D \operatorname{div}(\nabla u_0 - u_0 w \nabla \Phi), \quad \text{where } w = \beta \sum_{i=1}^n z_i u_i.$$

On the discrete level, we replace $u_0 w \nabla \Phi$ by an upwind approximation. This allows us to apply the discrete maximum principle showing that $u_0 \geq 0$ and hence $u = (u_1, \dots, u_n) \in \bar{\mathcal{D}}$ with \mathcal{D} defined in (6). Finally, Assumption (A3) is needed to derive a discrete version of the entropy inequality. Without the drift terms, the upwinding value does not depend on the index of the species, which simplifies some expressions; see Remark 2. \square

For the definition of the numerical scheme for (1)-(2), we need to introduce a suitable discretization of the domain Ω and the interval $(0, T)$. For simplicity, we consider a uniform

time discretization with time step $\Delta t > 0$, and we set $t^k = k\Delta t$ for $k = 1, \dots, N$, where $T > 0$, $N \in \mathbb{N}$ are given and $\Delta t = T/N$. The domain Ω is discretized by a regular and admissible triangulation in the sense of [16, Definition 9.1]. The triangulation consists of a family \mathcal{T} of open polygonal convex subsets of Ω (so-called cells), a family \mathcal{E} of edges (or faces in three dimensions), and a family of points $(x_K)_{K \in \mathcal{T}}$ associated to the cells. The admissibility assumption implies that the straight line between two centers of neighboring cells $\overline{x_K x_L}$ is orthogonal to the edge $\sigma = K|L$ between two cells K and L . The condition is satisfied by, for instance, triangular meshes whose triangles have angles smaller than $\pi/2$ [16, Examples 9.1] or Voronoi meshes [16, Example 9.2].

We assume that the family of edges \mathcal{E} can be split into internal and external edges $\mathcal{E} = \mathcal{E}_{\text{int}} \cup \mathcal{E}_{\text{ext}}$ with $\mathcal{E}_{\text{int}} = \{\sigma \in \mathcal{E} : \sigma \subset \Omega\}$ and $\mathcal{E}_{\text{ext}} = \{\sigma \in \mathcal{E} : \sigma \subset \partial\Omega\}$. Each exterior edge is assumed to be an element of either the Dirichlet or Neumann boundary, i.e. $\mathcal{E}_{\text{ext}} = \mathcal{E}_{\text{ext}}^D \cup \mathcal{E}_{\text{ext}}^N$. For given $K \in \mathcal{T}$, we define the set \mathcal{E}_K of the edges of K , which is the union of internal edges and edges on the Dirichlet or Neumann boundary, and we set $\mathcal{E}_{K,\text{int}} = \mathcal{E}_K \cap \mathcal{E}_{\text{int}}$.

The size of the mesh is defined by $h(\mathcal{T}) = \sup\{\text{diam}(K) : K \in \mathcal{T}\}$. For $\sigma \in \mathcal{E}_{\text{int}}$ with $\sigma = K|L$, we denote by $d_\sigma = d(x_K, x_L)$ the Euclidean distance between x_K and x_L , while for $\sigma \in \mathcal{E}_{\text{ext}}$, we set $d_\sigma = d(x_K, \sigma)$. For a given edge $\sigma \in \mathcal{E}$, the transmissibility coefficient is defined by

$$(9) \quad \tau_\sigma = \frac{m(\sigma)}{d_\sigma},$$

where $m(\sigma)$ denotes the Lebesgue measure of σ .

We impose a regularity assumption on the mesh: There exists $\zeta > 0$ such that for all $K \in \mathcal{T}$ and $\sigma \in \mathcal{E}_K$, it holds that

$$(10) \quad d(x_K, \sigma) \geq \zeta d_\sigma.$$

This hypothesis is needed to apply discrete functional inequalities (see [6, 16]) and a discrete compactness theorem (see [19]).

It remains to introduce suitable function spaces for the numerical discretization. The space $\mathcal{H}_{\mathcal{T}}$ of piecewise constant functions is defined by

$$\mathcal{H}_{\mathcal{T}} = \left\{ v : \overline{\Omega} \rightarrow \mathbb{R} : \exists (v_K)_{K \in \mathcal{T}} \subset \mathbb{R}, v(x) = \sum_{K \in \mathcal{T}} v_K \mathbf{1}_K(x) \right\}.$$

The (squared) discrete H^1 norm on this space is given by

$$(11) \quad \|v\|_{1,\mathcal{T}}^2 = \sum_{\sigma=K|L \in \mathcal{E}_{\text{int}}} \tau_\sigma (v_K - v_L)^2 + \sum_{K \in \mathcal{T}} m(K) v_K^2.$$

The discrete H^{-1} norm is the dual norm with respect to the L^2 scalar product,

$$(12) \quad \|v\|_{-1,\mathcal{T}} = \sup \left\{ \int_{\Omega} v w dx : w \in \mathcal{H}_{\mathcal{T}}, \|w\|_{1,\mathcal{T}} = 1 \right\}.$$

Then

$$\left| \int_{\Omega} v w dx \right| \leq \|v\|_{-1,\mathcal{T}} \|w\|_{1,\mathcal{T}} \quad \text{for } v, w \in \mathcal{H}_{\mathcal{T}}.$$

Finally, we introduce the space $\mathcal{H}_{\mathcal{T},\Delta t}$ of piecewise constant in time functions with values in $\mathcal{H}_{\mathcal{T}}$,

$$\mathcal{H}_{\mathcal{T},\Delta t} = \left\{ v : \bar{\Omega} \times [0, T] \rightarrow \mathbb{R} : \exists (v^k)_{k=1,\dots,N} \subset \mathcal{H}_{\mathcal{T}}, \quad v(x, t) = \sum_{k=1}^N v^k(x) \mathbf{1}_{(t^{k-1}, t^k)}(t) \right\},$$

equipped with the discrete $L^2(0, T; H^1(\Omega))$ norm

$$\|v\|_{1,\mathcal{T},\Delta t} = \left(\sum_{k=1}^N \Delta t \|v^k\|_{1,\mathcal{T}}^2 \right)^{1/2}.$$

For the numerical scheme, we introduce some further definitions. Let $u_i \in \mathcal{H}_{\mathcal{T}}$ with values $\bar{u}_{i,\sigma}$ on the Dirichlet boundary ($i = 1, \dots, n$). Then we introduce

$$(13) \quad D_{K,\sigma}(u_i) = u_{i,K,\sigma} - u_{i,K},$$

$$\text{where } u_{i,K,\sigma} = \begin{cases} u_{i,L} & \text{for } \sigma \in \mathcal{E}_{\text{int}}, \sigma = K|L, \\ \bar{u}_{i,\sigma} & \text{for } \sigma \in \mathcal{E}_{\text{ext},K}^D, \\ u_{i,K} & \text{for } \sigma \in \mathcal{E}_{\text{ext},K}^N, \end{cases} \quad \bar{u}_{i,\sigma} = \frac{1}{m(\sigma)} \int_{\sigma} \bar{u}_i ds.$$

The numerical fluxes $\mathcal{F}_{K,\sigma}$ should be consistent approximations to the exact fluxes through the edges $\int_{\sigma} \mathcal{F} \cdot \nu ds$. We impose the conservation of the numerical fluxes $\mathcal{F}_{K,\sigma} + \mathcal{F}_{L,\sigma} = 0$ for edges $\sigma = K|L$, requiring that they vanish on the Neumann boundary edges, $\mathcal{F}_{K,\sigma} = 0$ for $\sigma \in \mathcal{E}_{\text{ext},K}^N$. Then the discrete integration-by-parts formula becomes for $u \in \mathcal{H}_{\mathcal{T}}$

$$\sum_{K \in \mathcal{T}} \sum_{\sigma \in \mathcal{E}_K} \mathcal{F}_{K,\sigma} u_K = - \sum_{\sigma \in \mathcal{E}} \mathcal{F}_{K,\sigma} D_{K,\sigma}(u) + \sum_{\sigma \in \mathcal{E}_{\text{ext}}^D} \mathcal{F}_{K,\sigma} u_{K,\sigma}.$$

When $\partial\Omega = \Gamma_N$, this formula simplifies to

$$(14) \quad \sum_{K \in \mathcal{T}} \sum_{\sigma \in \mathcal{E}_K} \mathcal{F}_{K,\sigma} u_K = \sum_{\sigma = K|L \in \mathcal{E}_{\text{int}}} \mathcal{F}_{K,\sigma} (u_K - u_L).$$

2.2. Numerical scheme. We need to approximate the initial, boundary, and given functions on the elements $K \in \mathcal{T}$ and edges $\sigma \in \mathcal{E}$:

$$\begin{aligned} u_{i,K}^I &= \frac{1}{m(K)} \int_K u_i^I(x) dx, & f_K &= \frac{1}{m(K)} \int_K f(x) dx, \\ \bar{u}_{i,\sigma} &= \frac{1}{m(\sigma)} \int_{\sigma} \bar{u}_i ds, & \bar{\Phi}_{\sigma} &= \frac{1}{m(\sigma)} \int_{\sigma} \bar{\Phi} ds, \end{aligned}$$

and we set $u_{0,K}^I = 1 - \sum_{i=1}^n u_{i,K}^I$ and $\bar{u}_{0,\sigma} = 1 - \sum_{i=1}^n \bar{u}_{i,\sigma}$.

The numerical scheme is as follows. Let $K \in \mathcal{T}$, $k \in \{1, \dots, N\}$, $i = 1, \dots, n$, and $u_{i,K}^{k-1} \geq 0$ be given. Then the values $u_{i,K}^k$ are determined by the implicit Euler scheme

$$(15) \quad m(K) \frac{u_{i,K}^k - u_{i,K}^{k-1}}{\Delta t} + \sum_{\sigma \in \mathcal{E}_K} \mathcal{F}_{i,K,\sigma}^k = 0,$$

where the fluxes $\mathcal{F}_{i,K,\sigma}^k$ are given by the upwind scheme

$$(16) \quad \mathcal{F}_{i,K,\sigma}^k = -\tau_\sigma D_i \left(u_{0,\sigma}^k D_{K,\sigma}(u_i^k) - u_{i,\sigma}^k (D_{K,\sigma}(u_0^k) - \hat{u}_{0,\sigma,i}^k \beta z_i D_{K,\sigma}(\Phi^k)) \right),$$

where τ_σ is defined in (9),

$$(17) \quad u_{0,K}^k = 1 - \sum_{i=1}^n u_{i,K}^k, \quad u_{0,\sigma}^k = \max\{u_{0,K}^k, u_{0,L}^k\},$$

$$(18) \quad u_{i,\sigma}^k = \begin{cases} u_{i,K}^k & \text{if } \mathcal{V}_{i,K,\sigma}^k \geq 0, \\ u_{i,K,\sigma}^k & \text{if } \mathcal{V}_{i,K,\sigma}^k < 0, \end{cases}, \quad \hat{u}_{0,\sigma,i}^k = \begin{cases} u_{0,K}^k & \text{if } z_i D_{K,\sigma}(\Phi^k) \geq 0, \\ u_{0,K,\sigma}^k & \text{if } z_i D_{K,\sigma}(\Phi^k) < 0, \end{cases},$$

and $\mathcal{V}_{i,K,\sigma}^k$ is the “drift part” of the flux,

$$(19) \quad \mathcal{V}_{i,K,\sigma}^k = D_{K,\sigma}(u_0^k) - \hat{u}_{0,\sigma,i}^k \beta z_i D_{K,\sigma}(\Phi^k)$$

for $i = 1, \dots, n$. Observe that we employed a double upwinding: one related to the electric potential, defining $\hat{u}_{0,\sigma,i}^k$, and another one related to the drift part of the flux, $\mathcal{V}_{i,K,\sigma}^k$. The potential is computed via

$$(20) \quad -\lambda^2 \sum_{\sigma \in \mathcal{E}_K} \tau_\sigma D_{K,\sigma}(\Phi^k) = m(K) \left(\sum_{i=1}^n z_i u_{i,K}^k + f_K \right).$$

We recall that the numerical boundary conditions are given by $\bar{u}_{i,\sigma}$ and $\bar{\Phi}_\sigma$ for $\sigma \in \mathcal{E}_{\text{ext}}^D$.

We denote by $u_{i,\mathcal{T},\Delta t}$, $\Phi_{\mathcal{T},\Delta t}$ the functions in $\mathcal{H}_{\mathcal{T},\Delta t}$ associated to the values $u_{i,K}^k$ and Φ_K^k , respectively. Moreover, when dealing with a sequence of meshes $(\mathcal{T}_m)_m$ and a sequence of time steps $(\Delta t_m)_m$, we set $u_{i,m} = u_{i,\mathcal{T}_m,\Delta t_m}$, $\Phi_m = \Phi_{\mathcal{T}_m,\Delta t_m}$.

Remark 2 (Simplified numerical scheme). When Assumptions (A1)-(A3) hold, the numerical scheme simplifies to

$$(21) \quad m(K) \frac{u_{i,K}^k - u_{i,K}^{k-1}}{\Delta t} + \sum_{\sigma \in \mathcal{E}_{K,\text{int}}} \mathcal{F}_{i,K,\sigma}^k = 0,$$

$$(22) \quad \mathcal{F}_{i,K,\sigma}^k = -\tau_\sigma D \left(u_{0,\sigma}^k (u_{i,L}^k - u_{i,K}^k) - u_{i,\sigma}^k (u_{0,L}^k - u_{0,K}^k) \right),$$

where $u_{0,K}^k$ and $u_{0,\sigma}^k$ are defined in (17), and the definition of $u_{i,\sigma}^k$ simplifies to

$$u_{i,\sigma}^k = \begin{cases} u_{i,K}^k & \text{if } u_{0,K}^k - u_{0,L}^k \leq 0, \\ u_{i,L}^k & \text{if } u_{0,K}^k - u_{0,L}^k > 0. \end{cases}$$

In the definition of $u_{i,\sigma}^k$, the upwinding value does not depend on i anymore such that

$$(23) \quad \sum_{i=0}^n u_{i,\sigma}^k = 1 + \max\{u_{0,K}^k, u_{0,L}^k\} - \min\{u_{0,K}^k, u_{0,L}^k\} = 1 + |u_{0,K}^k - u_{0,L}^k|.$$

This property is needed to control the sum $\sum_{i=1}^n u_{i,\sigma}^k$ from below in the proof of the discrete entropy inequality; see (35). Finally, we are able to reformulate the discrete fluxes such that we obtain a discrete version of (8) (without the drift part):

$$(24) \quad \mathcal{F}_{i,K,\sigma} = \tau_\sigma D \left\{ u_{0,\sigma}^{1/2} (u_{0,K}^{1/2} u_{i,K} - u_{0,L}^{1/2} u_{i,L}) - u_{i,\sigma} (u_{0,K}^{1/2} - u_{0,L}^{1/2}) \left(u_{0,\sigma}^{1/2} + 2 \frac{u_{0,K}^{1/2} + u_{0,L}^{1/2}}{2} \right) \right\}.$$

This formulation is needed in the convergence analysis. \square

2.3. Main results. Since our scheme is implicit and nonlinear, the existence of an approximate solution is nontrivial. Therefore, our first result concerns the well-posedness of the numerical scheme.

Theorem 1 (Existence and uniqueness of solutions). *Let (H1)-(H4) and (A2) hold. Then there exists a solution (u, Φ) to scheme (15)-(20) satisfying $u^k \in \overline{\mathcal{D}}$ and, if the initial data lie in \mathcal{D} , $u^k \in \mathcal{D}$. If additionally Assumptions (A1) and (A3) hold, the solution is unique.*

Assumption (A2) is needed to show that $u_0^k = 1 - \sum_{i=1}^n u_i^k$ is nonnegative. Indeed, summing (15) and (16) over $i = 1, \dots, n$, we obtain

$$\mathfrak{m}(K) \frac{u_{0,K}^k - u_{0,K}^{k-1}}{\Delta t} = - \sum_{\sigma \in \mathcal{E}_K} \tau_\sigma \left(u_{0,\sigma}^k D_{K,\sigma} \left(\sum_{i=1}^n D_i u_i^k \right) - \sum_{i=1}^n D_i u_{i,\sigma}^k \mathcal{V}_{i,K,\sigma}^k \right).$$

Under Assumption (A2), it follows that $\sum_{i=1}^n D_i u_{i,K}^k = D(1 - u_{0,K}^k)$, and we can apply the discrete minimum principle, which then implies an L^∞ bound for u_i^k . This bound allows us to apply a topological degree argument; see [13, 14]. For the uniqueness proof, we additionally need Assumption (A3), since we use the entropy method of Gajewski [18], and it seems that this method cannot be applied to cross-diffusion systems including drift terms [28]. The idea is to prove first the uniqueness of u_0^k , which solves a discrete nonlinear equation, and then to show the uniqueness of u_i^k for $i = 1, \dots, n$ by introducing a semimetric $d(u^k, v^k)$ for two solutions $u^k = (u_1^k, \dots, u_n^k)$ and $v^k = (v_1^k, \dots, v_n^k)$ and showing that it is monotone in k , such that a discrete Gronwall argument implies that $u^k = v^k$.

The second result shows that the scheme preserves a discrete version of the entropy inequality.

Theorem 2 (Discrete entropy inequality). *Let Assumptions (H1)-(H4) and (A1)-(A3) hold. Then the solution to scheme (21)-(22) constructed in Theorem 1 satisfies the discrete entropy inequality*

$$(25) \quad \frac{H^k - H^{k-1}}{\Delta t} + I^k \leq 0,$$

with the discrete entropy

$$(26) \quad H^k = \sum_{K \in \mathcal{T}} m(K) \sum_{i=0}^n (u_{i,K}^k (\log u_{i,K}^k - 1) + 1)$$

and the discrete entropy production

$$\begin{aligned} I^k = D \sum_{\sigma=K|L \in \mathcal{E}_{\text{int}}} \tau_{\sigma} & \left(4 \sum_{i=1}^n u_{0,\sigma}^k ((u_{i,K}^k)^{1/2} - (u_{i,L}^k)^{1/2})^2 \right. \\ & \left. + 4((u_{0,K}^k)^{1/2} - (u_{0,L}^k)^{1/2})^2 + (u_{0,K}^k - u_{0,L}^k)^2 \right). \end{aligned}$$

Assumption (A3) is required to estimate the expression $\sum_{i=1}^n u_{i,\sigma}^k$. In the continuous case, this sum equals $1 - u_0$. On the discrete level, this identity cannot be expected since the value of $u_{i,\sigma}^k$ depends on the upwinding value; see (18). If the drift part vanishes, the upwinding value does not depend on i , as mentioned in Remark 2, and we can derive the estimate $\sum_{i=1}^n u_{i,\sigma}^k \geq 1 - u_{0,\sigma}^k$; see Section 4.1. Note that the entropy production I^k is the discrete counterpart of (7).

The main result of this paper is the convergence of the approximate solutions to a solution to the continuous cross-diffusion system.

Theorem 3 (Convergence of the approximate solution). *Let (H1)-(H4) and (A1)-(A3) hold and let (\mathcal{T}_m) and (Δt_m) be sequences of admissible meshes and time steps, respectively, such that $h(\mathcal{T}_m) \rightarrow 0$ and $\Delta t_m \rightarrow 0$ as $m \rightarrow \infty$. Let $(u_{0,m}, \dots, u_{n,m})$ be the solution to (21)-(22) constructed in Theorem 1. Then there exist functions $u_0, u = (u_1, \dots, u_n)$ satisfying $u(x, t) \in \overline{\mathcal{D}}$,*

$$\begin{aligned} u_0^{1/2}, u_0^{1/2} u_i & \in L^2(0, T; H^1(\Omega)), \quad i = 1, \dots, n, \\ u_{0,m}^{1/2} & \rightarrow u_0^{1/2}, \quad u_{0,m}^{1/2} u_{i,m} \rightarrow u_0^{1/2} u_i \quad \text{strongly in } L^2(\Omega \times (0, T)), \end{aligned}$$

where u is a weak solution to (1), (3)-(5) (with $\Gamma_N = \partial\Omega$), i.e., for all $\phi \in C_0^\infty(\overline{\Omega} \times [0, T])$ and $i = 1, \dots, n$,

$$(27) \quad \int_0^T \int_{\Omega} u_i \partial_t \phi dx dt + \int_{\Omega} u_i^I \phi(\cdot, 0) dx = D \int_0^T \int_{\Omega} u_0^{1/2} (\nabla(u_0^{1/2} u_i) - 3u_i \nabla u_0^{1/2}) \cdot \nabla \phi dx dt.$$

The compactness of the concentrations follows from the discrete gradient estimates derived from the entropy inequality (25), for which we need Assumption (A3). By the discrete Aubin-Lions lemma [17], we conclude the strong convergence of the sequence $(u_{0,m}^{1/2})$. The difficult part is to show the strong convergence of $(u_{0,m}^{1/2} u_{i,m})$, since there is no control on the discrete gradient of $u_{i,m}$. The idea is to apply a discrete Aubin-Lions lemma of “degenerate” type, proved in Lemma 10 in the appendix.

3. EXISTENCE AND UNIQUENESS OF APPROXIMATE SOLUTIONS

3.1. L^∞ bounds and existence of solutions. In order to prove the existence of solutions to (15)-(20), we first consider a truncated problem. This means that we truncate the expressions in (18); more precisely, we consider scheme (15), (16), and (20) with

$$(28) \quad \begin{aligned} u_{0,K}^k &= 1 - \sum_{i=1}^n (u_{i,K}^k)^+, \quad u_{0,\sigma}^k = \max\{0, u_{0,K}^k, u_{0,K,\sigma}^k\}, \\ \hat{u}_{0,\sigma,i}^k &= \begin{cases} (u_{0,K}^k)^+ & \text{if } z_i D_{K,\sigma}(\Phi^k) \geq 0, \\ (u_{0,K,\sigma}^k)^+ & \text{if } z_i D_{K,\sigma}(\Phi^k) < 0, \end{cases} \\ u_{i,\sigma}^k &= \begin{cases} (u_{i,K}^k)^+ & \text{if } \mathcal{V}_{i,K,\sigma}^k \geq 0, \\ (u_{i,K,\sigma}^k)^+ & \text{if } \mathcal{V}_{i,K,\sigma}^k < 0, \end{cases} \end{aligned}$$

where $z^+ = \max\{0, z\}$ for $z \in \mathbb{R}$ and $i = 1, \dots, n$. We show that this truncation is, in fact, not needed if the initial data are nonnegative. In the following let (H1)-(H4) hold.

Lemma 4 (Nonnegativity of u_i^k). *Let (u, Φ) be a solution to (15), (16), (20), and (28). Then $u_{i,K}^k \geq 0$ for all $K \in \mathcal{T}$, $k \in \{1, \dots, N\}$, and $i = 1, \dots, n$. If $u_i^I > 0$ and $\bar{u}_i > 0$ then also $u_{i,K}^k > 0$ for all $K \in \mathcal{T}$, $k \in \{1, \dots, N\}$.*

Proof. We proceed by induction. For $k = 0$, the nonnegativity holds because of our assumptions on the initial data. Assume that $u_{i,L}^{k-1} \geq 0$ for all $L \in \mathcal{T}$. Then let $u_{i,K}^k = \min\{u_{i,L}^k : L \in \mathcal{T}\}$ for some $K \in \mathcal{T}$ and assume that $u_{i,K}^k < 0$. The scheme writes as

$$(29) \quad m(K) \frac{u_{i,K}^k - u_{i,K}^{k-1}}{\Delta t} = \sum_{\sigma \in \mathcal{E}_K} \tau_\sigma D_i (u_{0,\sigma}^k D_{K,\sigma}(u_i^k) - u_{i,\sigma}^k \mathcal{V}_{i,K,\sigma}^k).$$

By assumption, $D_{K,\sigma}(u_i^k) \geq 0$. If $\mathcal{V}_{i,K,\sigma}^k \geq 0$, we have $-u_{i,\sigma}^k \mathcal{V}_{i,K,\sigma}^k = -(u_{i,K}^k)^+ \mathcal{V}_{i,K,\sigma}^k = 0$ and if $\mathcal{V}_{i,K,\sigma}^k < 0$, it follows that $-u_{i,\sigma}^k \mathcal{V}_{i,K,\sigma}^k = -(u_{i,K,\sigma}^k)^+ \mathcal{V}_{i,K,\sigma}^k \geq 0$. Hence, the right-hand side of (29) nonnegative. However, the left-hand side is negative, which is a contradiction. We infer that $u_{i,K}^k \geq 0$ and consequently, $u_{i,L}^k \geq 0$ for all $L \in \mathcal{T}$. When the initial data are positive, similar arguments show the positivity of $u_{i,L}^k$ for $L \in \mathcal{T}$. \square

We are able to show the nonnegativity of $u_{0,K}^k = 1 - \sum_{i=1}^n u_{i,K}^k$ only if the diffusion coefficients are the same. The reason is that we derive an equation for $u_{0,K}^k$ by summing (15) for $i = 1, \dots, n$, and this gives an equation for $u_{0,K}^k$ only if $D_i = D$ for all $i = 1, \dots, n$.

Lemma 5 (Nonnegativity of u_0^k). *Let Assumption (A2) hold and let (u, Φ) be a solution to (15), (16), (20), and (28). Then $u_{0,K}^k \geq 0$ for all $K \in \mathcal{T}$, $k \in \{1, \dots, N\}$. If $u_0^I > 0$ and $\bar{u}_i > 0$ then also $u_{0,K}^k > 0$ for all $K \in \mathcal{T}$, $k \in \{1, \dots, N\}$.*

Proof. Again, we proceed by induction. The case $k = 0$ follows from the assumptions. Assume that $u_{0,L}^{k-1} \geq 0$ for all $L \in \mathcal{T}$. Then let $u_{0,K}^k = \min\{u_{0,L}^k : L \in \mathcal{T}\}$ for some $K \in \mathcal{T}$

and assume that $u_{0,K}^k < 0$. Summing equations (15) from $i = 1, \dots, n$, we obtain

$$(30) \quad \begin{aligned} m(K) \frac{u_{0,K}^k - u_{0,K}^{k-1}}{\Delta t} &= D \sum_{\sigma \in \mathcal{E}_K} \tau_\sigma \left(u_{0,\sigma}^k D_{K,\sigma}(u_0^k) + \sum_{i=1}^n u_{i,\sigma}^k (D_{K,\sigma}(u_0^k) - \beta z_i \hat{u}_{0,\sigma,i}^k D_{K,\sigma}(\Phi^k)) \right) \\ &\geq -D \sum_{\sigma \in \mathcal{E}_K} \tau_\sigma \sum_{i=1}^n \beta z_i \hat{u}_{0,\sigma,i}^k D_{K,\sigma}(\Phi^k), \end{aligned}$$

since $u_{0,\sigma}^k \geq 0$ and $u_{i,\sigma}^k \geq 0$ by construction and $D_{K,\sigma}(u_0^k) \geq 0$ because of the minimality property of $u_{0,K}^k$. The remaining expression is nonnegative:

$$-\hat{u}_{0,\sigma,i}^k z_i D_{K,\sigma}(\Phi^k) = \begin{cases} -(u_{0,K}^k)^+ z_i D_{K,\sigma}(\Phi^k) = 0 & \text{if } z_i D_{K,\sigma}(\Phi^k) \geq 0, \\ -(u_{0,L}^k)^+ z_i D_{K,\sigma}(\Phi^k) \geq 0 & \text{if } z_i D_{K,\sigma}(\Phi^k) < 0. \end{cases}$$

However, the left-hand side of (30) is negative, by induction hypothesis, which gives a contradiction. \square

Lemmas 4 and 5 imply that we may remove the truncation in (28). Moreover, by definition, we have $1 - \sum_{i=1}^n u_{i,K}^k = u_{0,K}^k \geq 0$ such that $u_K^k = (u_{1,K}^k, \dots, u_{n,K}^k) \in \overline{\mathcal{D}}$ or, if the initial and boundary data are positive, $u_K^k \in \mathcal{D}$.

Proposition 6 (Existence for the numerical scheme). *Let Assumption (A2) hold. Then scheme (15)-(20) has a solution (u, Φ) which satisfies $u_K^k \in \overline{\mathcal{D}}$ for all $K \in \mathcal{T}$ and $k \in \mathbb{N}$.*

Proof. We argue by induction. For $k = 0$, we have $u_K^0 \in \overline{\mathcal{D}}$ by assumption. The function Φ^0 is uniquely determined by scheme (20), as this is a linear system of equations with positive definite matrix. Assume the existence of a solution (u^{k-1}, Φ^{k-1}) with $u_K^{k-1} \in \overline{\mathcal{D}}$. Let $m \in \mathbb{N}$ be the product of the number of species n and the number of cells $K \in \mathcal{T}$. For given $K \in \mathcal{T}$ and $i = 1, \dots, n$, we define the function $F_{i,K} : \mathbb{R}^m \times [0, 1] \rightarrow \mathbb{R}$ by

$$\begin{aligned} F_{i,K}(u, \rho) &= m(K) \frac{u_{i,K} - u_{i,K}^{k-1}}{\Delta t} \\ &\quad - \rho D \sum_{\sigma \in \mathcal{E}_K} \tau_\sigma \left(u_{0,\sigma} D_{K,\sigma}(u_i) - u_{i,\sigma} (D_{K,\sigma}(u_0) - \hat{u}_{0,\sigma,i} \beta z_i D_{K,\sigma}(\Phi)) \right). \end{aligned}$$

where $u_{0,K}$, $u_{i,\sigma}$, $u_{0,\sigma}$, and $\hat{u}_{0,\sigma,i}$ are defined in (28), and Φ is uniquely determined by (20). Let $F = (F_{i,K})_{i=1,\dots,n, K \in \mathcal{T}}$. Then $F : \mathbb{R}^m \times [0, 1] \rightarrow \mathbb{R}^m$ is a continuous function. We wish to apply the fixed-point theorem of [17, Theorem 5.1]. For this, we need to verify three assumptions:

- The function $u \mapsto F_{i,K}(u, 0) = m(K)(u_{i,K} - u_{i,K}^{k-1})/\Delta t$ is affine.
- We have proved above that any solution to $F(u, 1) = 0$ satisfies $u \in \mathcal{D}$ or $\|u\|_\infty < 2$. A similar proof shows that any solution to $F(u, \rho) = 0$ with $\rho \in (0, 1)$ satisfies $\|u\|_\infty < 2$, too.
- The equation $F(u, 0) = 0$ has the unique solution $u = u^{k-1}$ and consequently, $\|u\|_\infty = \|u^{k-1}\|_\infty < 2$.

We infer the existence of a solution u^k to $F(u^k, 1) = 0$ satisfying $\|u^k\|_\infty < 2$. In fact, by Lemmas 4 and 5, we find that $u^k \in \overline{\mathcal{D}}$. Hence, u^k solves the original scheme (15)-(20). \square

3.2. Uniqueness of solutions. The proof of Theorem 1 is completed when we show the uniqueness of solutions to scheme (15)-(20) under the additional conditions (A1) and (A3). Recall that in this case, the scheme is given by (21)-(22),

Step 1: uniqueness for u_0 . If $k = 0$, the solution is uniquely determined by the initial condition. Assume that u_0^{k-1} is given. Thanks to Assumptions (A2)-(A3), the sum of (21)-(22) for $i = 1, \dots, n$ gives an equation for $u_0^k = 1 - \sum_{i=1}^n u_i^k$ (in the following, we omit the superindices k):

$$\begin{aligned} m(K) \frac{u_{0,K} - u_{0,K}^{k-1}}{\Delta t} &= -D \sum_{\sigma \in \mathcal{E}_{K,\text{int}}} \tau_\sigma(u_{0,K} - u_{0,L}) \left(u_{0,\sigma} + \sum_{i=1}^n u_{i,\sigma} \right) \\ &= -D \sum_{\sigma \in \mathcal{E}_{K,\text{int}}} \tau_\sigma(u_{0,K} - u_{0,L}) (1 + |u_{0,K} - u_{0,L}|), \end{aligned}$$

where we used (23) in the last step.

Let u_0 and v_0 be two solutions to the previous equation and set $w_0 := u_0 - v_0$. Then w_0 solves

$$\begin{aligned} 0 &= m(K) \frac{w_{0,K}}{\Delta t} + D \sum_{\sigma \in \mathcal{E}_{K,\text{int}}} \tau_\sigma(w_{0,K} - w_{0,L}) \\ &\quad + D \sum_{\sigma \in \mathcal{E}_{K,\text{int}}} \tau_\sigma((u_{0,K} - u_{0,L})|u_{0,K} - u_{0,L}| - (v_{0,K} - v_{0,L})|v_{0,K} - v_{0,L}|). \end{aligned}$$

We multiply this equation by $w_{0,K}/D$, sum over $K \in \mathcal{T}$, and use discrete integration by parts (14):

$$\begin{aligned} 0 &= \sum_{K \in \mathcal{T}} \frac{m(K)}{D} \frac{w_{0,K}^2}{\Delta t} + \sum_{\sigma=K|L \in \mathcal{E}_{\text{int}}} \tau_\sigma(w_{0,K} - w_{0,L})^2 \\ &\quad + \sum_{\sigma=K|L \in \mathcal{E}_{\text{int}}} \tau_\sigma((u_{0,K} - u_{0,L})|u_{0,K} - u_{0,L}| - (v_{0,K} - v_{0,L})|v_{0,K} - v_{0,L}|)(w_{0,K} - w_{0,L}). \end{aligned}$$

The first two terms on the right-hand side are clearly nonnegative. We infer from the elementary inequality $(y|y| - z|z|)(y - z) \geq 0$ for $y, z \in \mathbb{R}$, which is a consequence of the monotonicity of $z \mapsto z|z|$, that the third term is nonnegative, too. Consequently, the three terms must vanish and this implies that $w_{0,K} = 0$ for all $K \in \mathcal{T}$. This shows the uniqueness for u_0 .

Step 2: uniqueness for u_i . Let u_0 be the uniquely determined solution from the previous step and let $u^k = (u_1^k, \dots, u_n^k)$ and $v^k = (v_1^k, \dots, v_n^k)$ be two solutions to (15). Similarly as in [18], we introduce the semimetric

$$d_\varepsilon(u^k, v^k) = \sum_{K \in \mathcal{T}} m(K) \sum_{i=1}^n H_1^\varepsilon(u_{i,K}^k, v_{i,K}^k), \quad \text{where}$$

$$H_1^\varepsilon(a, b) = h_\varepsilon(a) + h_\varepsilon(b) - 2h_\varepsilon\left(\frac{a+b}{2}\right)$$

and $h_\varepsilon(z) = (z + \varepsilon)(\log(z + \varepsilon) - 1) + 1$. The parameter $\varepsilon > 0$ is needed since $u_{i,K}^k$ or $v_{i,K}^k$ may vanish and then the logarithm of $u_{i,K}^k$ or $v_{i,K}^k$ may be undefined. The objective is to verify that $\lim_{\varepsilon \rightarrow 0} d_\varepsilon(u^k, v^k) = 0$ by estimating the discrete time derivative of the semimetric, implying that $u^k = v^k$.

First, we write

$$d_\varepsilon(u^k, v^k) - d_\varepsilon(u^{k-1}, v^{k-1}) = \sum_{K \in \mathcal{T}} m(K) \sum_{i=1}^n (H_1^\varepsilon(u_{i,K}^k, v_{i,K}^k) - H_1^\varepsilon(u_{i,K}^{k-1}, v_{i,K}^{k-1})).$$

The function H_1^ε is convex since

$$D^2 H_1^\varepsilon(a, b) = \frac{1}{(a + \varepsilon)(b + \varepsilon)(a + b + 2\varepsilon)} \begin{pmatrix} (b + \varepsilon)^2 & -(a + \varepsilon)(b + \varepsilon) \\ -(a + \varepsilon)(b + \varepsilon) & (a + \varepsilon)^2 \end{pmatrix}.$$

Therefore, a Taylor expansion of H_1^ε around $(u_{i,K}^k, v_{i,K}^k)$ leads to

$$\begin{aligned} & \frac{1}{\Delta t} (d_\varepsilon(u^k, v^k) - d_\varepsilon(u^{k-1}, v^{k-1})) \\ & \leq \sum_{K \in \mathcal{T}} \frac{m(K)}{\Delta t} \sum_{i=1}^n \left\{ D H_1^\varepsilon(u_{i,K}^k, v_{i,K}^k) \left(\begin{pmatrix} u_{i,K}^k \\ v_{i,K}^k \end{pmatrix} - \begin{pmatrix} u_{i,K}^{k-1} \\ v_{i,K}^{k-1} \end{pmatrix} \right) \right\} \\ & = \sum_{i=1}^n \sum_{K \in \mathcal{T}} m(K) \frac{u_{i,K}^k - u_{i,K}^{k-1}}{\Delta t} \left(h'_\varepsilon(u_{i,K}^k) - h'_\varepsilon\left(\frac{u_{i,K}^k + v_{i,K}^k}{2}\right) \right) \\ & \quad + \sum_{i=1}^n \sum_{K \in \mathcal{T}} m(K) \frac{v_{i,K}^k - v_{i,K}^{k-1}}{\Delta t} \left(h'_\varepsilon(v_{i,K}^k) - h'_\varepsilon\left(\frac{u_{i,K}^k + v_{i,K}^k}{2}\right) \right). \end{aligned}$$

We insert the scheme (21)-(22) and use discrete integration by parts:

$$\frac{1}{\Delta t} (d_\varepsilon(u^k, v^k) - d_\varepsilon(u^{k-1}, v^{k-1})) \leq S_1^k + S_2^k + \varepsilon S_3^k,$$

where

$$\begin{aligned} S_1^k &= -D \sum_{i=1}^n \sum_{\sigma=K | L \in \mathcal{E}_{\text{int}}} \tau_\sigma u_{0,\sigma}^k \left\{ (u_{i,K}^k - u_{i,L}^k) (\log(u_{i,K}^k + \varepsilon) - \log(u_{i,L}^k + \varepsilon)) \right. \\ & \quad + (v_{i,K}^k - v_{i,L}^k) (\log(v_{i,K}^k + \varepsilon) - \log(v_{i,L}^k + \varepsilon)) \\ & \quad \left. - 2 \left(\frac{u_{i,K}^k + v_{i,K}^k}{2} - \frac{u_{i,L}^k + v_{i,L}^k}{2} \right) \left(\log\left(\frac{u_{i,K}^k + v_{i,K}^k}{2} + \varepsilon\right) - \log\left(\frac{u_{i,L}^k + v_{i,L}^k}{2} + \varepsilon\right) \right) \right\}, \\ S_2^k &= D \sum_{i=1}^n \sum_{\sigma=K | L \in \mathcal{E}_{\text{int}}} \tau_\sigma (u_{0,K}^k - u_{0,L}^k) \left\{ (u_{i,\sigma}^k + \varepsilon) (\log(u_{i,K}^k + \varepsilon) - \log(u_{i,L}^k + \varepsilon)) \right. \\ & \quad \left. + (v_{i,\sigma}^k + \varepsilon) (\log(v_{i,K}^k + \varepsilon) - \log(v_{i,L}^k + \varepsilon)) \right\} \end{aligned}$$

$$\begin{aligned}
& -2 \left(\frac{u_{i,\sigma}^k + v_{i,\sigma}^k}{2} + \varepsilon \right) \left(\log \left(\frac{u_{i,K}^k + v_{i,K}^k}{2} + \varepsilon \right) - \log \left(\frac{u_{i,L}^k + v_{i,L}^k}{2} + \varepsilon \right) \right) \Bigg\}, \\
S_3^k &= -D \sum_{i=1}^n \sum_{\sigma=K|L \in \mathcal{E}_{\text{int}}} \tau_\sigma (u_{0,K}^k - u_{0,L}^k) \Bigg\{ (\log(u_{i,K}^k + \varepsilon) - \log(u_{i,L}^k + \varepsilon)) \\
& + (\log(v_{i,K}^k + \varepsilon) - \log(v_{i,L}^k + \varepsilon)) \\
& - 2 \left(\log \left(\frac{u_{i,K}^k + v_{i,K}^k}{2} + \varepsilon \right) - \log \left(\frac{u_{i,L}^k + v_{i,L}^k}{2} + \varepsilon \right) \right) \Bigg\}
\end{aligned}$$

We claim that $S_1^k \leq 0$ and $S_2^k \leq 0$. Indeed, with the definition $H_2^\varepsilon(a, b) = (a - b)(\log(a + \varepsilon) - \log(b + \varepsilon))$, we can reformulate S_1^k as

$$\begin{aligned}
S_1^k &= -D \sum_{i=1}^n \sum_{\sigma=K|L \in \mathcal{E}_{\text{int}}} \tau_\sigma u_{0,\sigma}^k \Bigg\{ H_2^\varepsilon(u_{i,K}^k, u_{i,L}^k) + H_2^\varepsilon(v_{i,K}^k, v_{i,L}^k) \\
& - 2H_2^\varepsilon\left(\frac{u_{i,K}^k + v_{i,K}^k}{2}, \frac{u_{i,L}^k + v_{i,L}^k}{2}\right) \Bigg\}.
\end{aligned}$$

The Hessian of H_2^ε ,

$$D^2 H_2^\varepsilon(a, b) = \begin{pmatrix} \frac{a+b+2\varepsilon}{(a+\varepsilon)^2} & -\frac{a+b+2\varepsilon}{(a+\varepsilon)(b+\varepsilon)} \\ -\frac{a+b+2\varepsilon}{(a+\varepsilon)(b+\varepsilon)} & \frac{a+b+2\varepsilon}{(b+\varepsilon)^2} \end{pmatrix},$$

is positive semidefinite. Therefore, performing a Taylor expansion up to second order, we see that $S_1^k \leq 0$.

Next, we show that $S_2^k \leq 0$. For this, we assume without loss of generality for some fixed $\sigma = K|L$ that $u_{0,K}^k \leq u_{0,L}^k$. By definition of the scheme, $u_{i,\sigma}^k = u_{i,K}^k$ and $v_{i,\sigma}^k = v_{i,K}^k$. Set $H_3^\varepsilon(a, b) = (a + \varepsilon)(\log(a + \varepsilon) - \log(b + \varepsilon))$. The term in the curly bracket in S_2^k then takes the form

$$(31) \quad (u_{0,K}^k - u_{0,L}^k) \Bigg\{ H_3^\varepsilon(u_{i,K}^k, u_{i,L}^k) + H_3^\varepsilon(v_{i,K}^k, v_{i,L}^k) - 2H_3^\varepsilon\left(\frac{u_{i,K}^k + v_{i,K}^k}{2}, \frac{u_{i,L}^k + v_{i,L}^k}{2}\right) \Bigg\}.$$

The Hessian of H_3^ε ,

$$D^2 H_3^\varepsilon(a, b) = \begin{pmatrix} \frac{1}{a+\varepsilon} & -\frac{1}{b+\varepsilon} \\ -\frac{1}{b+\varepsilon} & \frac{1}{b+\varepsilon} \end{pmatrix},$$

is also positive semidefinite, showing that (31) is nonpositive as $u_{0,K}^k - u_{0,L}^k \leq 0$. If $u_{0,K}^k > u_{0,L}^k$, both factors of the product (31) change their sign, so that we arrive at the same conclusion. Hence, $S_2^k \leq 0$. We conclude that

$$d_\varepsilon(u^k, v^k) - d_\varepsilon(u^{k-1}, v^{k-1}) \leq \varepsilon \Delta t S_3^k.$$

Since $d_\varepsilon(u^0, v^0) = 0$, we find after resolving the recursion that

$$d_\varepsilon(u^k, v^k) \leq \varepsilon \Delta t \sum_{\ell=1}^k S_3^\ell.$$

As the densities $u_{i,K}^\ell$ are nonnegative and bounded by 1 for all $K \in \mathcal{T}$, for all $\ell \geq 0$ and for all $1 \leq i \leq n$, it is clear that $\sum_{\ell=1}^k \varepsilon S_3^\ell \rightarrow 0$ when $\varepsilon \rightarrow 0$. Then, we may perform the limit $\varepsilon \rightarrow 0$ in the previous inequality yielding $d_\varepsilon(u^k, v^k) \rightarrow 0$. A Taylor expansion as in [28, end of Section 6] shows that $d_\varepsilon(u^k, v^k) \geq \frac{1}{8} \sum_{K \in \mathcal{T}} m(K) \sum_{i=1}^n (u_{i,K}^k - v_{i,K}^k)^2$. We infer that $u^k = v^k$, finishing the proof.

4. DISCRETE ENTROPY INEQUALITY AND UNIFORM ESTIMATES

4.1. Discrete entropy inequality. First, we prove (25).

Proof of Theorem 2. The idea is to multiply (15) by $\log(u_{i,K}^{k,\varepsilon}/u_{0,K}^{k,\varepsilon})$, where $u_{i,K}^{k,\varepsilon} := u_{i,K}^k + \varepsilon$ for $i = 0, \dots, n$. The regularization is necessary to avoid issues when the concentrations vanish. After this multiplication, we sum the equations over $i = 1, \dots, n$ and $K \in \mathcal{T}$ and use discrete integration by parts to obtain

$$\begin{aligned}
 0 &= \sum_{K \in \mathcal{T}} \frac{m(K)}{\Delta t D} \sum_{i=1}^n (u_{i,K}^k - u_{i,K}^{k-1}) \log \frac{u_{i,K}^{k,\varepsilon}}{u_{0,K}^{k,\varepsilon}} \\
 (32) \quad &+ \sum_{\sigma=K|L \in \mathcal{E}_{\text{int}}} \tau_\sigma \left(u_{0,\sigma}^k (u_{i,K}^k - u_{i,L}^k) - u_{i,\sigma}^k (u_{0,K}^k - u_{0,L}^k) \right) \left(\log \frac{u_{i,K}^{k,\varepsilon}}{u_{0,K}^{k,\varepsilon}} - \log \frac{u_{i,L}^{k,\varepsilon}}{u_{0,L}^{k,\varepsilon}} \right) \\
 &= A_0 + \sum_{\sigma=K|L \in \mathcal{E}_{\text{int}}} \tau_\sigma (A_1 + A_2 + B_1 + B_2),
 \end{aligned}$$

where

$$\begin{aligned}
 A_0 &= \sum_{K \in \mathcal{T}} \frac{m(K)}{\Delta t D} \sum_{i=0}^n (u_{i,K}^{k,\varepsilon} - u_{i,K}^{k-1,\varepsilon}) \log u_{i,K}^{k,\varepsilon}, \\
 A_1 &= \sum_{i=1}^n u_{0,\sigma}^k (u_{i,K}^{k,\varepsilon} - u_{i,L}^{k,\varepsilon}) (\log u_{i,K}^{k,\varepsilon} - \log u_{i,L}^{k,\varepsilon}), \\
 A_2 &= - \sum_{i=1}^n u_{0,\sigma}^k (u_{i,K}^k - u_{i,L}^k) (\log u_{0,K}^{k,\varepsilon} - \log u_{0,L}^{k,\varepsilon}), \\
 B_1 &= - \sum_{i=1}^n u_{i,\sigma}^k (u_{0,K}^k - u_{0,L}^k) (\log u_{i,K}^{k,\varepsilon} - \log u_{i,L}^{k,\varepsilon}), \\
 B_2 &= \sum_{i=1}^n u_{i,\sigma}^k (u_{0,K}^k - u_{0,L}^k) (\log u_{0,K}^{k,\varepsilon} - \log u_{0,L}^{k,\varepsilon}).
 \end{aligned}$$

The convexity of $h(z) = z(\log z - 1) + 1$ implies the inequality $h(u) - h(v) \leq h'(u)(u - v)$ for all $u, v \in \mathbb{R}$. Consequently,

$$A_0 \geq \sum_{K \in \mathcal{T}} \frac{m(K)}{\Delta t D} \sum_{i=0}^n (u_{i,K}^{k,\varepsilon} (\log u_{i,K}^{k,\varepsilon} - 1) - u_{i,K}^{k-1,\varepsilon} (\log u_{i,K}^{k-1,\varepsilon} - 1)).$$

In order to estimate the remaining terms, we recall two elementary inequalities. Let $y, z > 0$. Then, by the Cauchy-Schwarz inequality,

$$(33) \quad (\sqrt{y} - \sqrt{z})^2 = \left(\int_z^y \frac{ds}{2\sqrt{s}} \right)^2 \leq \int_z^y \frac{ds}{4} \int_z^y \frac{ds}{s} = \frac{1}{4}(y - z)(\log y - \log z),$$

and by the concavity of the logarithm,

$$(34) \quad y(\log y - \log z) \geq y - z \geq z(\log y - \log z).$$

Inequality (33) shows that

$$A_1 \geq 4 \sum_{i=1}^n u_{0,\sigma}^k ((u_{i,K}^{k,\varepsilon})^{1/2} - (u_{i,K}^{k,\varepsilon})^{1/2}).$$

We use the definition of $u_{0,K}^k = 1 - \sum_{i=1}^n u_{i,K}^k$ in A_2 to find that

$$A_2 = u_{0,\sigma}^k (u_{0,K}^k - u_{0,L}^k) (\log u_{0,K}^{k,\varepsilon} - \log u_{0,L}^{k,\varepsilon}).$$

We rewrite B_1 by using the abbreviation $u_{i,\sigma}^{k,\varepsilon} = u_{i,\sigma}^k + \varepsilon$:

$$\begin{aligned} B_1 &= - \sum_{i=1}^n u_{i,\sigma}^{k,\varepsilon} (u_{0,K}^k - u_{0,L}^k) (\log u_{i,K}^{k,\varepsilon} - \log u_{i,L}^{k,\varepsilon}) \\ &\quad + \varepsilon \sum_{i=1}^n (u_{0,K}^k - u_{0,L}^k) (\log u_{i,K}^{k,\varepsilon} - \log u_{i,L}^{k,\varepsilon}) \\ &=: B_{11} + \varepsilon B_{12}. \end{aligned}$$

We apply inequality (34) to B_{11} . Indeed, if $u_{0,K}^k \leq u_{0,L}^k$, we have $u_{i,\sigma}^k = u_{i,K}^k$ and we use the first inequality in (34). If $u_{0,K}^k > u_{0,L}^k$ then $u_{i,\sigma}^k = u_{i,L}^k$ and we employ the second inequality in (34). In both cases, it follows that

$$\begin{aligned} B_{11} &\geq - \sum_{i=1}^n (u_{0,K}^k - u_{0,L}^k) (u_{i,K}^{k,\varepsilon} - u_{i,L}^{k,\varepsilon}) \\ &= - (u_{0,K}^k - u_{0,L}^k) \sum_{i=1}^n (u_{i,K}^{k,\varepsilon} - u_{i,L}^{k,\varepsilon}) = (u_{0,K}^k - u_{0,L}^k)^2. \end{aligned}$$

Finally, we consider B_2 . In view of Assumption (A3), equation (23) gives

$$(35) \quad \sum_{i=1}^n u_{i,\sigma}^k = 1 - \min\{u_{0,K}^k, u_{0,L}^k\} \geq 1 - u_{0,\sigma}^k,$$

and therefore, by (33),

$$\begin{aligned} B_2 &\geq (1 - u_{0,\sigma}^k) (u_{0,K}^{k,\varepsilon} - u_{0,L}^{k,\varepsilon}) (\log u_{0,K}^{k,\varepsilon} - \log u_{0,L}^{k,\varepsilon}) \\ &\geq 4((u_{0,K}^{k,\varepsilon})^{1/2} - (u_{0,L}^{k,\varepsilon})^{1/2})^2 - u_{0,\sigma}^k (u_{0,K}^k - u_{0,L}^k) (\log u_{0,K}^{k,\varepsilon} - \log u_{0,L}^{k,\varepsilon}). \end{aligned}$$

The last expression cancels with A_2 such that

$$A_2 + B_2 \geq 4((u_{0,K}^{k,\varepsilon})^{1/2} - (u_{0,L}^{k,\varepsilon})^{1/2})^2.$$

Putting together the estimates for A_0 , A_1 , B_1 , and $A_2 + B_2$, we deduce from (32) that

$$\begin{aligned}
& \sum_{K \in \mathcal{T}} \frac{m(K)}{\Delta t} \sum_{i=0}^n u_{i,K}^{k,\varepsilon} (\log u_{i,K}^{k,\varepsilon} - 1) - \sum_{K \in \mathcal{T}} \frac{m(K)}{\Delta t} \sum_{i=1}^n u_{i,K}^{k-1,\varepsilon} (\log u_{i,K}^{k-1,\varepsilon} - 1) \\
& + D \sum_{\sigma=K|L \in \mathcal{E}_{\text{int}}} \tau_\sigma \left\{ 4 \sum_{i=1}^n u_{0,\sigma} ((u_{i,K}^{k,\varepsilon})^{1/2} - (u_{i,L}^{k,\varepsilon})^{1/2})^2 \right. \\
& \left. + 4((u_{0,K}^{k,\varepsilon})^{1/2} - (u_{0,L}^{k,\varepsilon})^{1/2})^2 + (u_{0,K}^k - u_{0,L}^k)^2 \right\} \\
& \leq -\varepsilon D (u_{0,K}^k - u_{0,L}^k) \sum_{i=1}^n (\log u_{i,K}^{k,\varepsilon} - \log u_{i,L}^{k,\varepsilon}).
\end{aligned}$$

Since the right-hand side converges to zero as $\varepsilon \rightarrow 0$, we infer that (25) holds. \square

4.2. A priori estimates. For the proof of the convergence result, we need estimates uniform in the mesh size $h(\mathcal{T})$ and time step Δt . The scheme provides uniform L^∞ bounds. Further bounds are derived from the discrete entropy inequality of Theorem 2. We introduce the discrete time derivative for functions $v \in \mathcal{H}_{\mathcal{T},\Delta t}$ by

$$(36) \quad \partial_t^{\Delta t} v^k = \frac{v^k - v^{k-1}}{\Delta t}, \quad k = 1, \dots, N.$$

Lemma 7 (A priori estimates). *Let (H1)-(H4) and (A1)-(A3) hold. The solution u to scheme (21)-(22) satisfies the following uniform estimates:*

$$(37) \quad \|u_0^{1/2}\|_{1,\mathcal{T},\Delta t} + \|u_0^{1/2} u_i\|_{1,\mathcal{T},\Delta t} \leq C, \quad i = 1, \dots, n,$$

$$(38) \quad \sum_{k=1}^N \Delta t \|\partial_t^{\Delta t} u_i^k\|_{-1,\mathcal{T}}^2 \leq C, \quad i = 0, \dots, n,$$

where the constant $C > 0$ is independent of the mesh \mathcal{T} and time step size Δt .

Proof. We claim that estimates (37) follow from the discrete entropy inequality (25). Indeed, we sum (25) over $k = 1, \dots, N$ to obtain

$$\begin{aligned}
& H^N + D \sum_{k=1}^N \Delta t \sum_{\sigma=K|L \in \mathcal{E}_{\text{int}}} \tau_\sigma \left(4 \sum_{i=1}^n u_{0,\sigma}^k ((u_{i,K}^k)^{1/2} - (u_{i,L}^k)^{1/2})^2 \right. \\
& \left. + 4((u_{0,K}^k)^{1/2} - (u_{0,L}^k)^{1/2})^2 + (u_{0,K}^k - u_{0,L}^k)^2 \right) \leq H^0.
\end{aligned}$$

Since the entropy at time $t = 0$ is bounded independently of the discretization, we infer immediately the bound for $u_0^{1/2}$ in $\mathcal{H}_{\mathcal{T},\Delta t}$. For the bound on $u_0^{1/2} u_i$ in $\mathcal{H}_{\mathcal{T},\Delta t}$, we observe that

$$\begin{aligned}
& (u_{0,K}^k)^{1/2} u_{i,K}^k - (u_{0,L}^k)^{1/2} u_{i,L}^k \\
& = u_{i,K} ((u_{0,K}^k)^{1/2} - (u_{0,L}^k)^{1/2}) + (u_{0,L}^k)^{1/2} ((u_{i,K}^k)^{1/2} + (u_{i,L}^k)^{1/2}) ((u_{i,K}^k)^{1/2} - (u_{i,L}^k)^{1/2}).
\end{aligned}$$

Therefore, together with the L^∞ bounds on u_i ,

$$\begin{aligned} & \sum_{\sigma=K|L \in \mathcal{E}_{\text{int}}} \tau_\sigma \left((u_{0,K}^k)^{1/2} u_{i,K}^k - (u_{0,L}^k)^{1/2} u_{i,L}^k \right)^2 \\ & \leq \sum_{\sigma=K|L \in \mathcal{E}_{\text{int}}} \tau_\sigma \left((u_{0,K}^k)^{1/2} - (u_{0,L}^k)^{1/2} \right)^2 + 2 \sum_{\sigma=K|L \in \mathcal{E}_{\text{int}}} \tau_\sigma u_{0,\sigma}^k \left((u_{i,K}^k)^{1/2} - (u_{i,L}^k)^{1/2} \right)^2. \end{aligned}$$

Then, summing over $k = 0, \dots, N$ and using the estimates from the entropy inequality, we achieve the bound on $u_0^{1/2} u_i$.

It remains to prove estimate (38). To this end, let $\phi \in \mathcal{H}_{\mathcal{T}}$ be such that $\|\phi\|_{1,\mathcal{T}} = 1$ and let $k \in \{1, \dots, N\}$ and $i \in \{1, \dots, n\}$. We multiply the scheme (21) by Φ_K and we sum over $K \in \mathcal{T}$. Using successively discrete integration by parts, the rewriting of the numerical fluxes (24), the Cauchy-Schwarz inequality, and the L^∞ bounds on u_i , we compute

$$\begin{aligned} & \sum_{K \in \mathcal{T}} \frac{m(K)}{\Delta t} (u_{i,K}^k - u_{i,K}^{k-1}) \phi_K \\ & = D \sum_{\sigma=K|L \in \mathcal{E}_{\text{int}}} \tau_\sigma (u_{0,\sigma}^k)^{1/2} \left((u_{0,K}^k)^{1/2} u_{i,K}^k - (u_{0,L}^k)^{1/2} u_{i,L}^k \right) (\phi_K - \phi_L) \\ & \quad - D \sum_{\sigma=K|L \in \mathcal{E}_{\text{int}}} \tau_\sigma \left((u_{0,K}^k)^{1/2} - (u_{0,L}^k)^{1/2} \right) \\ & \quad \times u_{i,\sigma}^k \left((u_{0,\sigma}^k)^{1/2} + 2 \frac{(u_{0,K}^k)^{1/2} + (u_{0,L}^k)^{1/2}}{2} \right) (\phi_K - \phi_L) \\ & \leq D \left(\sum_{\sigma=K|L \in \mathcal{E}_{\text{int}}} \tau_\sigma \left((u_{0,K}^k)^{1/2} u_{i,K}^k - (u_{0,L}^k)^{1/2} u_{i,L}^k \right)^2 \right)^{1/2} \left(\sum_{\sigma=K|L \in \mathcal{E}_{\text{int}}} \tau_\sigma (\phi_K - \phi_L)^2 \right)^{1/2} \\ & \quad + 3D \left(\sum_{\sigma=K|L \in \mathcal{E}_{\text{int}}} \tau_\sigma \left((u_{0,K}^k)^{1/2} - (u_{0,L}^k)^{1/2} \right)^2 \right)^{1/2} \left(\sum_{\sigma=K|L \in \mathcal{E}_{\text{int}}} \tau_\sigma (\phi_K - \phi_L)^2 \right)^{1/2}. \end{aligned}$$

This shows that, for $i = 1, \dots, n$,

$$\sum_{k=1}^N \Delta t \left\| \frac{u_i^k - u_i^{k-1}}{\Delta t} \right\|_{-1,\mathcal{T}}^2 \leq 2D^2 \sum_{k=1}^N \Delta t \left(\|(u_0^k)^{1/2} u_i^k\|_{1,\mathcal{T}}^2 + 9 \|(u_0^k)^{1/2}\|_{1,\mathcal{T}}^2 \right) \leq C,$$

as a consequence of (37). The estimate for $\Delta t^{-1} (u_0^k - u_0^{k-1}) = -\Delta t^{-1} \sum_{i=1}^n (u_i^k - u_i^{k-1})$ follows from those for $i = 1, \dots, n$, completing the proof. \square

5. CONVERGENCE OF THE SCHEME

In this section, we establish the convergence of the sequence of approximate solutions, constructed in Theorem 1, to a weak solution to (1), i.e., we prove Theorem 3.

5.1. Compactness of the approximate solutions. In order to achieve the convergence in the fluxes, we proceed as in [12] by defining the approximate gradient on a dual mesh. For $\sigma = K|L \in \mathcal{E}_{\text{int}}$, we define the new cell T_{KL} as the cell with the vertexes x_K , x_L and those of σ . For $\sigma \in \mathcal{E}_{\text{ext}} \cap \mathcal{E}_K$, we define $T_{K\sigma}$ as the cell with vertex x_K and those of σ . Then Ω can be decomposed as

$$\bar{\Omega} = \bigcup_{K \in \mathcal{T}} \left\{ \left(\bigcup_{L \in \mathcal{N}_K} \bar{T}_{KL} \right) \cup \left(\bigcup_{\sigma \in \mathcal{E}_{\text{ext}, K}} \bar{T}_{K\sigma} \right) \right\},$$

where \mathcal{N}_K denotes the set of neighboring cells of K . The discrete gradient $\nabla_{\mathcal{T}, \Delta t} v$ on $\Omega_T := \Omega \times (0, T)$ for piecewise constant functions $v \in \mathcal{H}_{\mathcal{T}, \Delta t}$ is defined by

$$(39) \quad \nabla_{\mathcal{T}, \Delta t} v(x, t) = \begin{cases} \frac{m(\sigma)(v_L^k - v_K^k)}{m(T_{KL})} \mathbf{n}_{KL} & \text{for } x \in T_{KL}, \quad t \in (t^k, t^{k+1}), \\ 0 & \text{for } x \in T_{K\sigma}, \quad t \in (t^k, t^{k+1}), \end{cases}$$

where \mathbf{n}_{KL} denotes the unit normal on $\sigma = K|L$ oriented from K to L . To simplify the notation, we set $\nabla_m := \nabla_{\mathcal{T}_m, \Delta t_m}$. The solution to the approximate scheme (21)-(22) is called $u_{0,m}, u_{1,m}, \dots, u_{n,m}$.

Lemma 8. *There exist functions $u_0 \in L^\infty(\Omega_T) \cap L^2(0, T; H^1(\Omega))$ and $u_1, \dots, u_n \in L^\infty(\Omega_T)$ such that, possibly for subsequences, as $m \rightarrow \infty$,*

$$(40) \quad u_{0,m} \rightarrow u_0, \quad u_{0,m}^{1/2} \rightarrow u_0^{1/2} \quad \text{strongly in } L^2(\Omega_T),$$

$$(41) \quad \nabla_m u_{0,m} \rightharpoonup \nabla u_0, \quad \nabla_m u_{0,m}^{1/2} \rightharpoonup \nabla u_0^{1/2} \quad \text{weakly in } L^2(\Omega_T),$$

$$(42) \quad u_{0,m}^{1/2} u_{i,m} \rightarrow u_0^{1/2} u_i \quad \text{strongly in } L^2(\Omega_T),$$

$$(43) \quad \nabla_m (u_{0,m}^{1/2} u_{i,m}) \rightharpoonup \nabla (u_0^{1/2} u_i) \quad \text{weakly in } L^2(\Omega_T),$$

where $i \in \{1, \dots, n\}$.

Proof. First, we claim that $(u_{0,m})$ is uniformly bounded in $\mathcal{H}_{\mathcal{T}, \Delta t}$. Indeed, by the L^∞ bounds and estimate (37),

$$\begin{aligned} \|u_{0,m}\|_{1, \mathcal{T}, \Delta t}^2 &= \sum_{k=1}^N \Delta t \left(\sum_{\sigma=K|L \in \mathcal{E}_{\text{int}}} \tau_\sigma (u_{0,K}^k - u_{0,L}^k)^2 + \sum_{K \in \mathcal{T}} m(K) (u_{0,K}^k)^2 \right) \\ &= \sum_{k=1}^N \Delta t \left(\sum_{\sigma=K|L \in \mathcal{E}_{\text{int}}} \tau_\sigma ((u_{0,K}^k)^{1/2} + (u_{0,L}^k)^{1/2})^2 ((u_{0,K}^k)^{1/2} - (u_{0,L}^k)^{1/2})^2 \right. \\ &\quad \left. + \sum_{K \in \mathcal{T}} m(K) (u_{0,K}^k)^2 \right) \\ &\leq 4 \|u_{0,m}\|_{L^\infty(\Omega_T)} \|u_{0,m}^{1/2}\|_{1, \mathcal{T}, \Delta t}^2 + \|u_{0,m}\|_{L^2(\Omega_T)}^2 \leq C. \end{aligned} \quad (44)$$

By estimate (38), $(\partial_t^{\Delta t} u_{0,m})$ is uniformly bounded. Therefore, by the discrete Aubin-Lions lemma (see Lemma 9 in the appendix), we conclude the existence of a subsequence (not

reabeled) such that the first convergence in (40) holds. The strong convergence implies (up to a subsequence) that $u_{0,m} \rightarrow u_0$ pointwise in Ω_T and consequently $u_{0,m}^{1/2} \rightarrow u_0^{1/2}$ pointwise in Ω_T . Thus, together with the L^∞ bound for $u_{0,m}^{1/2}$, we infer the second convergence in (40).

The convergences in (41) are a consequence of the uniform estimates (37) and (44) and the compactness result in [16, proof of Theorem 10.3]. Applying the discrete Aubin-Lions lemma of “degenerate” type (see Lemma 10 in the appendix) to $y_m = u_{0,m}^{1/2}$ and $z_m = u_{i,m}$ for fixed $i \in \{1, \dots, n\}$, we deduce convergence (42). Finally, convergence (43) is a consequence of (42) and the weak compactness of $(u_{0,m}^{1/2} u_{i,m})$, thanks to the uniform bound in (37). \square

5.2. The limit $m \rightarrow \infty$. We finish the proof of Theorem 3 by verifying that the limit function $u = (u_1, \dots, u_n)$, as defined in Lemma 8, is a weak solution in the sense of the theorem.

Let $\phi \in C_0^\infty(\bar{\Omega} \times [0, T))$ and let $m \in \mathbb{N}$ be large enough such that $\text{supp } \phi \subset \bar{\Omega} \times [0, (N_m - 1)\Delta t_m)$ (recall that $T = N_m \Delta t_m$). For the limit, we follow the strategy used, for instance, in [12] and introduce the following notations:

$$\begin{aligned} F_{10}(m) &= - \int_0^T \int_\Omega u_{i,m} \partial_t \phi dx dt - \int_\Omega u_{i,m}(0) \phi(0) dx, \\ F_{20}(m) &= \int_0^T \int_\Omega u_{0,m}^{1/2} \nabla_m(u_{0,m}^{1/2} u_{i,m}) \nabla \phi dx dt, \\ F_{30}(m) &= 3 \int_0^T \int_\Omega u_{0,m}^{1/2} u_{i,m} \nabla_m(u_{0,m}^{1/2}) \nabla \phi dx dt. \end{aligned}$$

The convergence results of Lemma 8 show that, as $m \rightarrow \infty$,

$$\begin{aligned} (45) \quad F_{10}(m) + DF_{20}(m) - DF_{30}(m) &\rightarrow - \int_0^T \int_\Omega u_i \partial_t \phi dx dt - \int_\Omega u_i^0 \phi(0) dx \\ &\quad + D \int_0^T \int_\Omega (u_0^{1/2} \nabla(u_0^{1/2} u_i) - 3u_0^{1/2} u_i \nabla u_0^{1/2}) dx dt. \end{aligned}$$

Next, setting $\phi_K^k = \phi(x_K, t^k)$, we multiply scheme (21) by $\Delta t_m \phi_K^{k-1}$ and sum over $K \in \mathcal{T}_m$ and $k = 1, \dots, N_m$. Then

$$(46) \quad F_1(m) + DF_2(m) - DF_3(m) = 0,$$

where, omitting the subscript m from now on to simplify the notation,

$$\begin{aligned} F_1(m) &= \sum_{k=1}^N \sum_{K \in \mathcal{T}} m(K) (u_{i,K}^k - u_{i,K}^{k-1}) \phi_K^{k-1}, \\ F_2(m) &= \sum_{k=1}^N \Delta t \sum_{K \in \mathcal{T}} \sum_{\sigma \in \mathcal{E}_{K,\text{int}}} \tau_\sigma (u_{0,\sigma}^k)^{1/2} ((u_{0,K}^k)^{1/2} u_{i,K}^k - (u_{0,L}^k)^{1/2} u_{i,L}^k) \phi_K^{k-1}, \end{aligned}$$

$$F_3(m) = \sum_{k=1}^N \Delta t \sum_{K \in \mathcal{T}} \sum_{\sigma \in \mathcal{E}_{K,\text{int}}} \tau_\sigma \left((u_{0,K}^k)^{1/2} - (u_{0,L}^k)^{1/2} \right) \\ \times u_{i,\sigma}^k \left((u_{0,\sigma}^k)^{1/2} + 2 \frac{(u_{0,K}^k)^{1/2} + (u_{0,L}^k)^{1/2}}{2} \right) \phi_K^{k-1}.$$

The aim is to show that $F_{i0}(m) - F_i(m) \rightarrow 0$ as $m \rightarrow \infty$ for $i = 1, 2, 3$. Then, because of (46), $F_{10}(m) + DF_{20}(m) - DF_{30}(m) \rightarrow 0$, which finishes the proof. We start by verifying that $F_{10}(m) - F_1(m) \rightarrow 0$. For this, we rewrite $F_1(m)$ and $F_{10}(m)$, using $\phi_K^N = 0$:

$$F_1(m) = \sum_{k=1}^N \sum_{K \in \mathcal{T}} m(K) u_{i,K}^k (\phi_K^{k-1} - \phi_K^k) - \sum_{K \in \mathcal{T}} m(K) \phi_K^0 u_{i,K}^0, \\ = - \sum_{k=1}^N \sum_{K \in \mathcal{T}} \int_{t^{k-1}}^{t^k} \int_K u_{i,K}^k \partial_t \phi(x_K, t) dx dt - \sum_{K \in \mathcal{T}} \int_K u_{i,K}^0 \phi(x_K, 0) dx, \\ F_{10}(m) = - \sum_{k=1}^N \sum_{K \in \mathcal{T}} \int_{t^{k-1}}^{t^k} \int_K u_{i,K}^k \partial_t \phi(x, t) dx dt - \sum_{K \in \mathcal{T}} \int_K u_{i,K}^0 \phi(x, 0) dx.$$

In view of the regularity of ϕ and the uniform L^∞ bound on u_i , we find that

$$|F_{10}(m) - F_1(m)| \leq CT m(\Omega) \|\phi\|_{C^2} h(\mathcal{T}_m) \rightarrow 0 \quad \text{as } m \rightarrow \infty.$$

Using discrete integration by parts, the second integral becomes

$$F_2(m) = \sum_{k=1}^N \Delta t \sum_{\sigma=K|L \in \mathcal{E}_{\text{int}}} \tau_\sigma (u_{0,\sigma}^k)^{1/2} \left((u_{0,K}^k)^{1/2} u_{i,K}^k - (u_{0,L}^k)^{1/2} u_{i,L}^k \right) (\phi_K^{k-1} - \phi_L^{k-1}) \\ = F_{21}(m) + F_{22}(m),$$

where we have decomposed $(u_{0,\sigma}^k)^{1/2} = (u_{0,K}^k)^{1/2} + ((u_{0,\sigma}^k)^{1/2} - (u_{0,K}^k)^{1/2})$, i.e.

$$F_{21}(m) = \sum_{k=1}^N \Delta t \sum_{\sigma=K|L \in \mathcal{E}_{\text{int}}} \tau_\sigma (u_{0,K}^k)^{1/2} \left((u_{0,K}^k)^{1/2} u_{i,K}^k - (u_{0,L}^k)^{1/2} u_{i,L}^k \right) (\phi_K^{k-1} - \phi_L^{k-1}), \\ F_{22}(m) = \sum_{k=1}^N \Delta t \sum_{\sigma=K|L \in \mathcal{E}_{\text{int}}} \tau_\sigma \left((u_{0,\sigma}^k)^{1/2} - (u_{0,K}^k)^{1/2} \right) \left((u_{0,K}^k)^{1/2} u_{i,K}^k - (u_{0,L}^k)^{1/2} u_{i,L}^k \right) \\ \times (\phi_K^{k-1} - \phi_L^{k-1}).$$

Furthermore, we write $F_{20}(m) = G_1(m) + G_2(m)$, where

$$G_1(m) = \sum_{k=1}^N \sum_{\sigma=K|L \in \mathcal{E}_{\text{int}}} \frac{m(\sigma)}{m(T_{KL})} (u_{0,K}^k)^{1/2} \left((u_{0,K}^k)^{1/2} u_{i,K}^k - (u_{0,L}^k)^{1/2} u_{i,L}^k \right) \\ \times \int_{t^{k-1}}^{t^k} \int_{T_{KL}} \nabla \phi(x, t) \cdot \mathbf{n}_{K\sigma} dx dt,$$

$$\begin{aligned}
G_2(m) &= \sum_{k=1}^N \sum_{\sigma=K|L \in \mathcal{E}_{\text{int}}} \frac{m(\sigma)}{m(T_{KL})} ((u_{0,L}^k)^{1/2} - (u_{0,K}^k)^{1/2}) ((u_{0,K}^k)^{1/2} u_{i,K}^k - (u_{0,L}^k)^{1/2} u_{i,L}^k) \\
&\quad \times \int_{t^{k-1}}^{t^k} \int_{T_{KL} \cap L} \nabla \phi(x, t) \cdot \mathbf{n}_{K\sigma} dx dt.
\end{aligned}$$

The aim is to show that $F_{21}(m) - G_1(m) \rightarrow 0$, $F_{22}(m) \rightarrow 0$, and $G_2(m) \rightarrow 0$. This implies that

$$\begin{aligned}
|F_{20}(m) - F_2(m)| &= |(G_1(m) + G_2(m)) - (F_{21}(m) + F_{22}(m))| \\
&\leq |G_1 - F_{21}| + |G_2| + |F_{22}| \rightarrow 0.
\end{aligned}$$

First we notice that, due to the admissibility of the mesh and the regularity of ϕ , by taking the mean value over T_{KL} ,

$$(47) \quad \left| \int_{t^{k-1}}^{t^k} \left(\frac{\phi_K^{k-1} - \phi_L^{k-1}}{d_\sigma} - \frac{1}{m(T_{KL})} \int_{T_{KL}} \nabla \phi(x, t) \cdot \mathbf{n}_{K\sigma} \right) dt \right| \leq C \Delta t h(\mathcal{T}),$$

where the constant $C > 0$ only depends on ϕ . It yields

$$\begin{aligned}
|F_{21}(m) - G_1(m)| &\leq Ch(\mathcal{T}) \sum_{k=1}^N \Delta t \sum_{\sigma=K|L \in \mathcal{E}_{\text{int}}} m(\sigma) |(u_{0,K}^k)^{1/2} u_{i,K}^k - (u_{0,L}^k)^{1/2} u_{i,L}^k| \\
&\leq Ch(\mathcal{T}) \|u_0^{1/2} u_i\|_{1, \mathcal{T}, \Delta t} (Tm(\Omega))^{1/2},
\end{aligned}$$

where the last estimate follows from the Cauchy-Schwarz inequality. This proves that $|F_{21}(m) - G_1(m)| \rightarrow 0$ as $m \rightarrow \infty$.

It remains to analyze the expressions $F_{22}(m)$ and $G_2(m)$. To this end, we remark that $d_\sigma \leq h(\mathcal{T})$ and hence, together with the regularity of ϕ , and the Cauchy-Schwarz inequality,

$$\begin{aligned}
|F_{22}(m)| &\leq \sum_{k=1}^N \Delta t \sum_{\sigma=K|L \in \mathcal{E}_{\text{int}}} \tau_\sigma |(u_{0,\sigma}^k)^{1/2} - (u_{0,K}^k)^{1/2}| |(u_{0,K}^k)^{1/2} u_{i,K}^k - (u_{0,L}^k)^{1/2} u_{i,L}^k| \\
&\quad \times \frac{|\phi_K^{k-1} - \phi_L^{k-1}|}{d_\sigma} d_\sigma \\
&\leq Ch(\mathcal{T}) \|\phi\|_{C^1} \sum_{k=1}^N \Delta t \sum_{\sigma=K|L \in \mathcal{E}_{\text{int}}} \tau_\sigma |(u_{0,\sigma}^k)^{1/2} - (u_{0,K}^k)^{1/2}| \\
&\quad \times |(u_{0,K}^k)^{1/2} u_{i,K}^k - (u_{0,L}^k)^{1/2} u_{i,L}^k| \\
&\leq Ch(\mathcal{T}) \|\phi\|_{C^1} \|u_0^{1/2}\|_{1, \mathcal{T}, \Delta t} \|u_0^{1/2} u_i\|_{1, \mathcal{T}, \Delta t} \leq Ch(\mathcal{T}),
\end{aligned}$$

The term $G_2(m)$ can be estimated in a similar way.

Finally, we need to show that $|F_{30}(m) - F_3(m)| \rightarrow 0$. The proof is completely analogous to the previous arguments, since

$$\left| 3(u_{0,K}^k)^{1/2} u_{i,K}^k - u_{i,\sigma}^k \left((u_{0,\sigma}^k)^{1/2} + 2 \frac{(u_{0,K}^k)^{1/2} + (u_{0,L}^k)^{1/2}}{2} \right) \right|$$

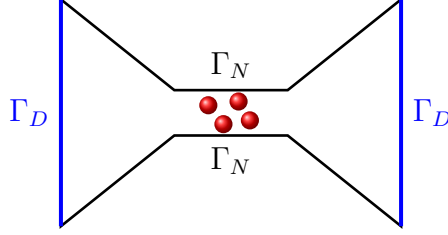


FIGURE 1. Schematic picture of the ion channel Ω used for the simulations. Dirichlet boundary conditions are prescribed on Γ_D (blue), homogeneous Neumann boundary conditions on Γ_N (black). The red circles represent the confined $O^{1/2-}$ ions.

$$\leq C((u_{0,\sigma}^k)^{1/2}|u_{i,K}^k - u_{i,L}^k| + |(u_{0,K}^k)^{1/2} - (u_{0,L}^k)^{1/2}|).$$

Summarizing, we have proved that $|F_{i0}(m) - F_i(m)| \rightarrow 0$ for $i = 1, 2, 3$, and since $F_1(m) + DF_2(m) - DF_3(m) = 0$, the convergence (45) shows that u solves (27). This completes the proof of Theorem 3.

6. NUMERICAL EXPERIMENTS

We present numerical simulations of a calcium-selective ion channel in two space dimensions to illustrate the dynamical behavior of the ion transport model. Numerical simulations in one space dimension can be found in [11] for stationary solutions and in [20] for transient solutions. The channel is modeled as in [21]. The selectivity of the channel is obtained by placing some confined oxygen ions ($O^{1/2-}$) inside the channel region. These ions contribute to the permanent charge density $f = -u_{\text{ox}}/2$ in the Poisson equation, but also to the total sum of the concentrations. We consider three further types of ions: calcium (Ca^{2+} , u_1), sodium (Na^+ , u_2), and chloride (Cl^- , u_3). While the concentrations of these ion species satisfy the evolution equations (1), the oxygen concentration is constant in time and given by the piecewise linear function

$$u_{\text{ox}}(x, y) = u_{\text{ox,max}} \times \begin{cases} 1 & \text{for } 0.45 \leq x \leq 0.55, \\ 10(x - 0.35) & \text{for } 0.35 \leq x \leq 0.45, \\ 10(0.65 - x) & \text{for } 0.55 \leq x \leq 0.65, \\ 0 & \text{else,} \end{cases}$$

where the scaled maximal oxygen concentration equals $u_{\text{ox,max}} = (N_A/u_{\text{typ}}) \cdot 52 \text{ mol/L}$, where $N_A \approx 6.022 \cdot 10^{23} \text{ mol}^{-1}$ is the Avogadro constant and $u_{\text{typ}} = 3.7037 \cdot 10^{25} L^{-1}$ the typical concentration (taken from [11, Table 1]). This gives $u_{\text{ox,max}} \approx 0.84$. The solvent concentration is computed according to $u_0 = 1 - \sum_{i=1}^3 u_i - u_{\text{ox}}$. The physical parameters used in our simulations are taken from [11, Table 1], and the channel geometry is depicted in Figure 1. The boundary conditions are chosen as in [11, Section 5].

The simulations are performed with the full set of equations (1)-(2) without assuming (A1)-(A3). The finite-volume scheme (15)-(20) is implemented using MATLAB, version

R2015a. The nonlinear system defined by the implicit scheme is solved with a full Newton method in the variables u_0, u_1, u_2, u_3, Φ for every time step. The computations are done with a fixed time step size $\Delta t = 10^{-3}$ until the stationary state is approximately reached, i.e., until the discrete L^2 norm between the solutions at two consecutive time steps is smaller than 10^{-12} . We employ an admissible mesh with 4736 elements generated by the MATLAB command `initmesh`, which produces Delauney meshes. As initial data, piecewise linear functions that connect the boundary values are chosen for the ion concentrations, while the initial potential is computed from the Poisson equation using the initial concentrations as charge density.

Figures 2 and 3 show the concentration profiles and the electric potential after 50 and 1400 time steps, respectively. The equilibrium is approximately reached after 1653 time steps. The profiles depicted in Figure 3 are already very close to the stationary state and correspond qualitatively well to the one-dimensional stationary profiles presented in [11]. We observe that during the evolution, sodium inside the channel is replaced by the stronger positively charged calcium ions. For higher initial calcium concentrations, the calcium selectivity of the channel acts immediately.

The simulations suggest that the solution tends towards a steady state as $t \rightarrow \infty$. The large-time behavior can be quantified by computing the relative entropy E^k with respect to the stationary solution, where

$$E^k = \sum_{K \in \mathcal{T}} m(K) \sum_{i=0}^n u_{i,K}^k \log \left(\frac{u_{i,K}^k}{u_{i,K}^\infty} \right) + \frac{\lambda^2}{2} \sum_{\sigma \in \mathcal{E}} \tau_\sigma D_{K,\sigma} (\Phi^k - \Phi^\infty)^2$$

and $(u_{i,K}^\infty, \Phi^\infty)$ is the constant steady state determined from the boundary data. Figure 4 shows that the relative entropy as well as the discrete L^1 norms of the concentrations and electric potential decay with exponential rate. Interestingly, after some initial phase, the convergence is rather slow and increases after this intermediate phase. This phase can be explained by the degeneracy at $u_0 = 0$, which causes a small entropy production slowing down diffusion. Indeed, as shown in [20] for the one-dimensional setting, a small change in the oxygen concentration may prolong the intermediate phase of slow convergence drastically.

Since Assumptions (A1)-(A3) are not satisfied in our test case, the convergence result of Theorem 3 cannot be applied here. However, we still observe convergence of the numerical solutions. As the exact solution is not known explicitly, we compute a reference solution on a very fine mesh with 75 776 elements and mesh size $h(\mathcal{T}) \approx 0.01$. This mesh is obtained from the coarse mesh by a regular refinement, dividing the triangles into four triangles of the same shape. The reference solution is compared to approximate solutions on coarser nested meshes. In Figure 5, the errors in the discrete L^1 norm between the reference solution and the solutions on the coarser meshes at two fixed time steps $k = 50$ and $k = 1400$ are plotted. We clearly observe the expected first-order convergence in space.

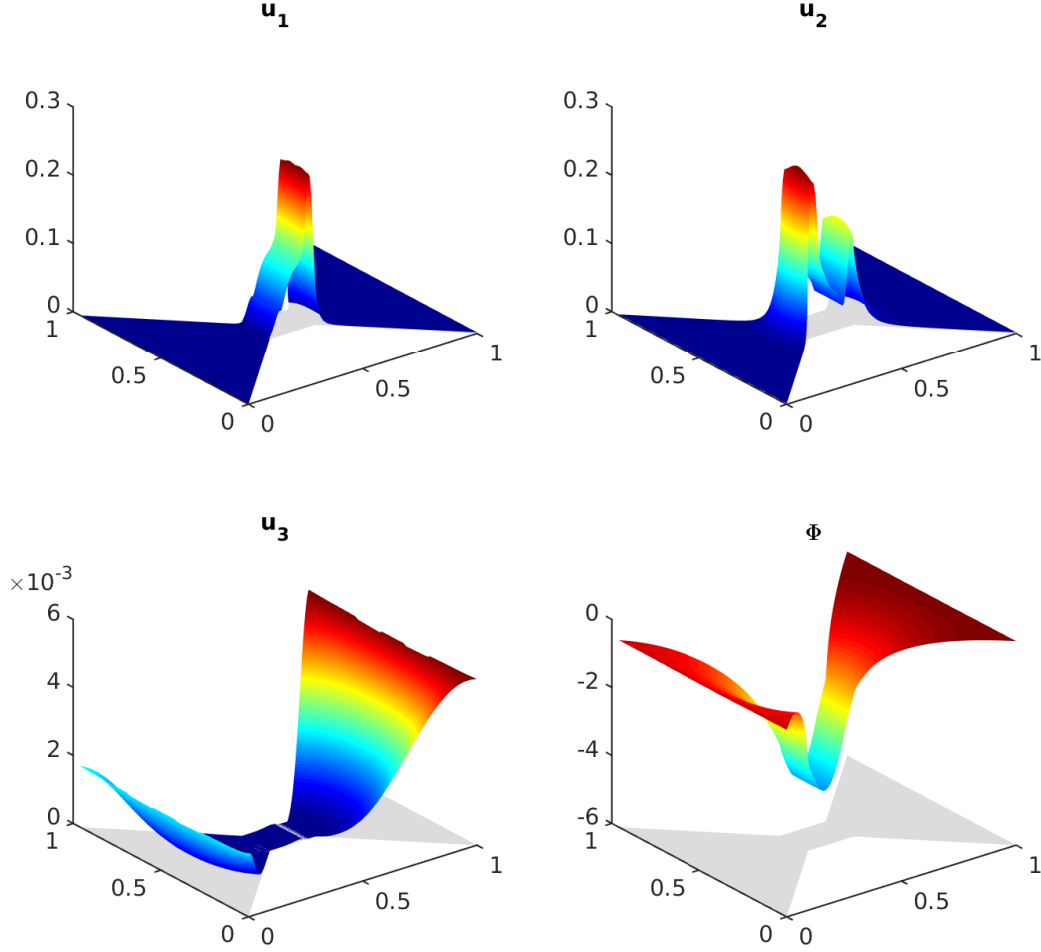


FIGURE 2. Scaled concentrations of calcium, sodium, and chloride ions and electric potential after 50 time steps.

APPENDIX A. AUXILIARY RESULTS

We prove two versions of discrete Aubin-Lions lemmas. The first one is a consequence of [19, Theorem 3.4], the second one extends Lemma 13 in [22] to the discrete case. The latter result is new. Recall that $\Omega_T = \Omega \times (0, T)$, $\nabla_m = \nabla_{\mathcal{T}_m, \Delta t_m}$ is the discrete gradient defined in (39), and $\partial_t^{\Delta t}$ is the discrete time derivative defined in (36).

Lemma 9 (Discrete Aubin-Lions). *Let $\|\cdot\|_{1, \mathcal{T}_m}$ be the norm on $\mathcal{H}_{\mathcal{T}_m}$ defined in (11) with the dual norm $\|\cdot\|_{-1, \mathcal{T}_m}$ given by (12), and let $(u_m) \subset \mathcal{H}_{\mathcal{T}_m, \Delta t_m}$ be a sequence of piecewise*

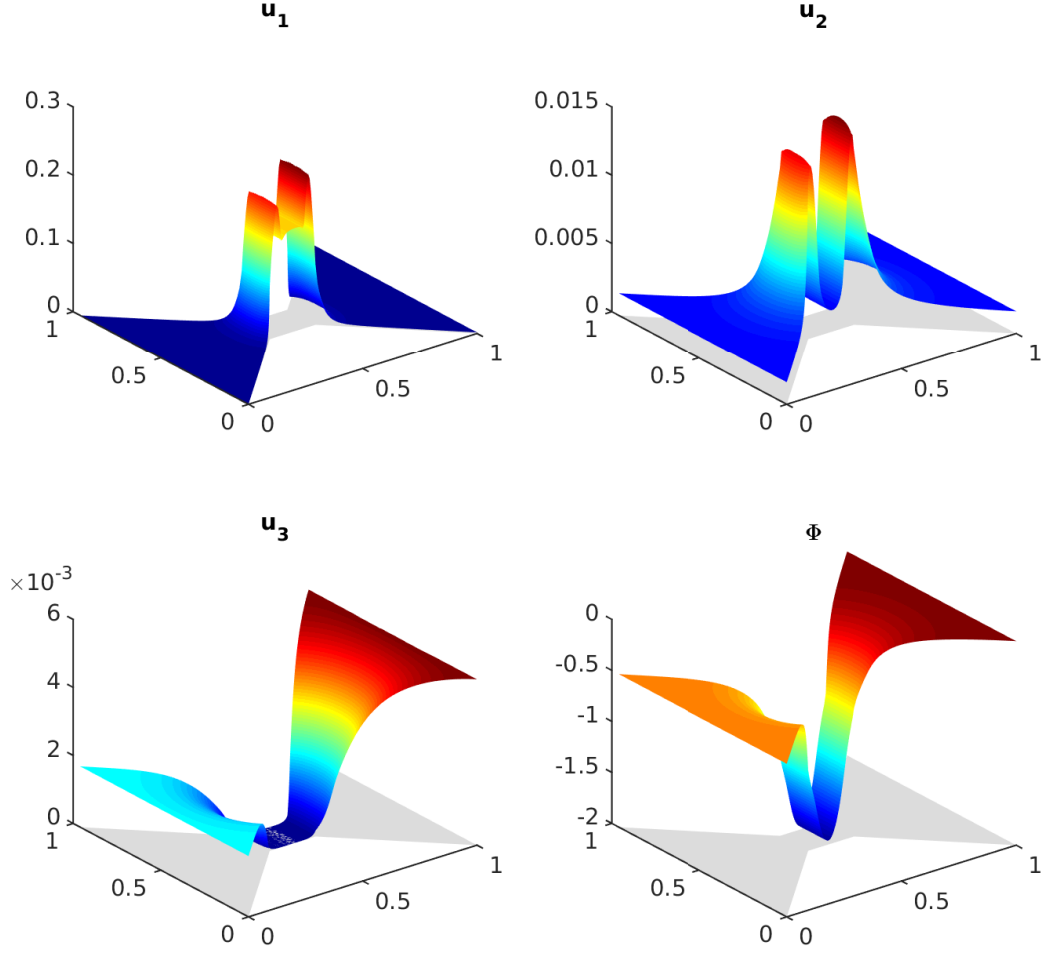


FIGURE 3. Scaled concentrations of calcium, sodium, and chloride ions and electric potential after 1400 time steps (close to equilibrium).

constants in time functions with values in $\mathcal{H}_{\mathcal{T}_m}$ satisfying

$$\sum_{k=1}^{N_m} \Delta t (\|u_m^k\|_{1,\mathcal{T}_m}^2 + \|\partial_t^{\Delta t_m} u_m^k\|_{-1,\mathcal{T}_m}^2) \leq C,$$

where $C > 0$ is independent of the size of the mesh and the time step size. Then there exists a subsequence, which is not relabeled, such that, as $m \rightarrow \infty$,

$$\begin{aligned} u_m &\rightarrow u \quad \text{strongly in } L^2(\Omega_T), \\ \nabla_m u_m &\rightharpoonup \nabla u \quad \text{weakly in } L^2(\Omega_T). \end{aligned}$$

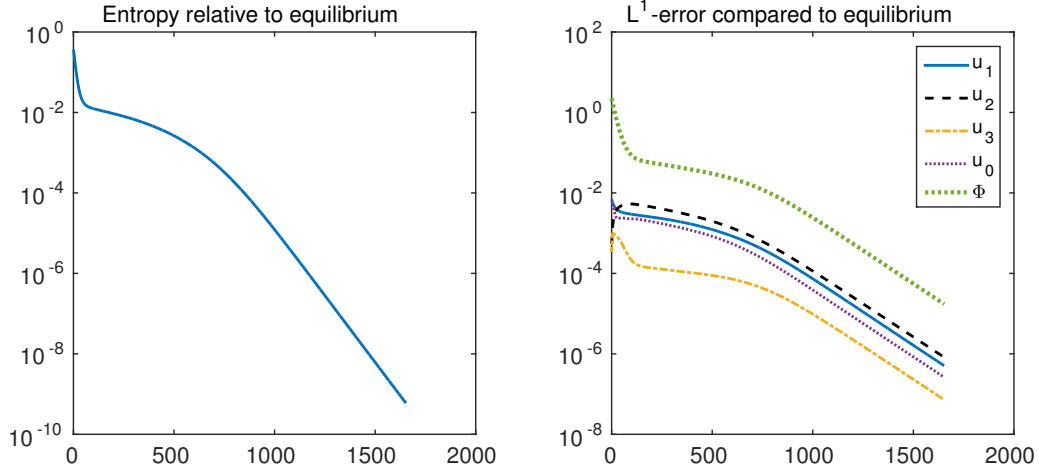


FIGURE 4. Relative entropy (left) and discrete L^1 error relative to the equilibrium (right) over the number of time steps.

Proof. The result is a consequence of Theorem 3.4 in [19]. To apply this theorem, we have to show that the discrete norms $\|\cdot\|_{1,\mathcal{T}_m}$ and $\|\cdot\|_{-1,\mathcal{T}_m}$ satisfy the assumptions of Lemma 3.1 in [19]:

- (1) For any sequence $(v_m) \subset \mathcal{H}_{\mathcal{T}_m}$ such that there exists $C > 0$ with $\|v_m\|_{1,\mathcal{T}_m} \leq C$ for all $m \in \mathbb{N}$, there exists $v \in L^2(\Omega)$ such that, up to a subsequence, $v_m \rightarrow v$ in $L^2(\Omega)$.
- (2) If $v_m \rightarrow v$ strongly in $L^2(\Omega)$ and $\|v_m\|_{-1,\mathcal{T}_m} \rightarrow 0$ as $m \rightarrow \infty$, then $v = 0$.

The first property is proved in, for instance, [17, Lemma 5.6]. Here, we need assumption (10) on the mesh. The second property can be replaced, according to [19, Remark 6], by the condition that $\|\cdot\|_{1,\mathcal{T}_m}$ and $\|\cdot\|_{-1,\mathcal{T}_m}$ are dual norms with respect to the $L^2(\Omega)$ norm, which is the case here. We infer that there exists a subsequence of (u_m) , which is not relabeled, such that $u_m \rightarrow u$ strongly in $L^2(\Omega_T)$. The weak convergence of the discrete gradients can be proved as in Lemma 4.4 in [12]. Indeed, the boundedness of $(\nabla_m u_m)$ in L^2 implies the convergence to some function $\chi \in L^2(\Omega_T)$ (up to a subsequence). In order to show that $\chi = \nabla u$, it remains to verify that for all test functions $\phi \in C_0^\infty(\Omega_T; \mathbb{R}^d)$,

$$\int_0^T \int_\Omega \nabla_m u_m \cdot \phi dx dt + \int_0^T \int_\Omega u_m \operatorname{div} \phi dx dt \rightarrow 0 \quad \text{as } m \rightarrow \infty.$$

This limit follows from the definition of $\nabla_m u_m$ and the regularity of the mesh. We refer to [12, Lemma 4.4] for details. \square

Lemma 10 (Discrete Aubin-Lions of “degenerate” type). *Let (y_m) and (z_m) be sequences in $\mathcal{H}_{\mathcal{T}_m, \Delta t_m}$ which are bounded in $L^\infty(\Omega_T)$ and let (y_m) be relatively compact in $L^2(\Omega_T)$, i.e., up to a subsequence, $y_m \rightarrow y$ strongly in $L^2(\Omega_T)$ and $z_m \rightharpoonup^* z$ weakly* in $L^\infty(\Omega_T)$.*

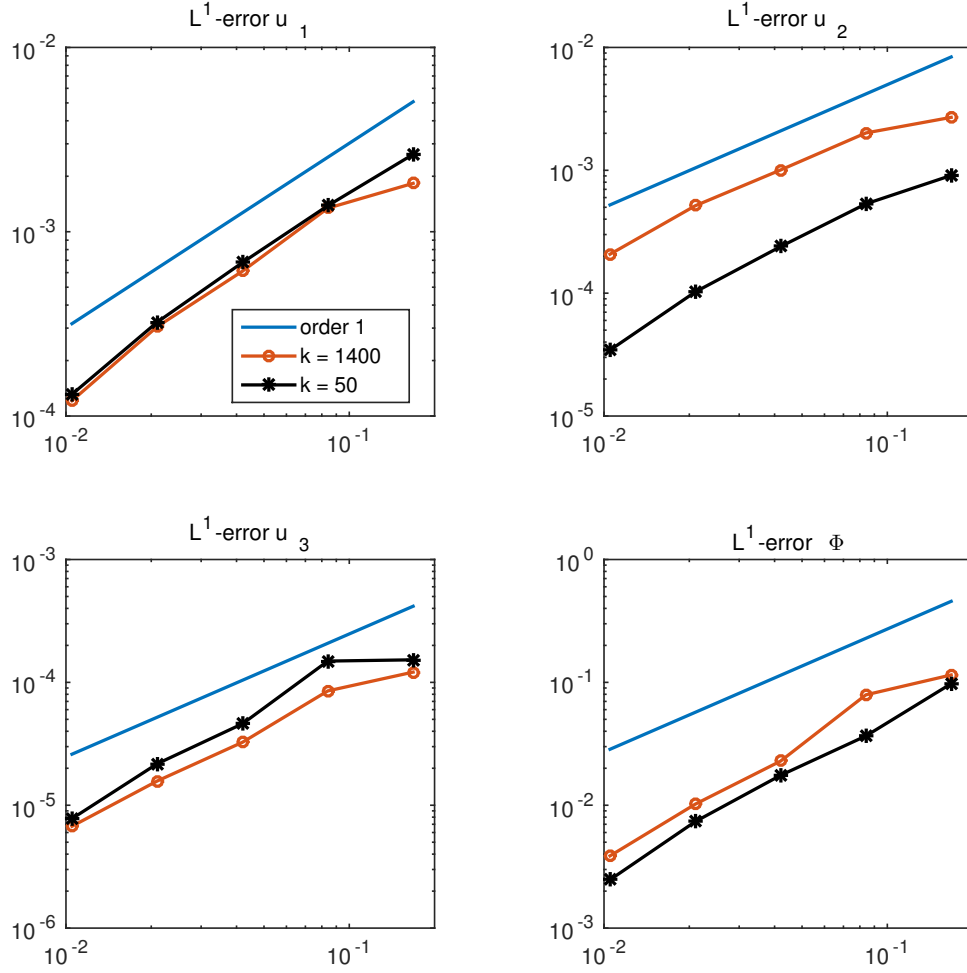


FIGURE 5. Discrete L^1 error relative to the reference solution at two different time steps over the mesh size $h(\mathcal{T})$.

Furthermore, suppose that, for some constant $C > 0$ independent of m ,

$$\sum_{k=1}^{N_m} \Delta t_m (\|y_m^k\|_{1,\mathcal{T}_m}^2 + \|y_m^k z_m^k\|_{1,\mathcal{T}_m}^2 + \|\partial_t^{\Delta t_m} z_m^k\|_{-1,\mathcal{T}_m}^2) \leq C.$$

Then there exists a subsequence which is not relabeled such that $y_m z_m \rightarrow yz$ strongly in $L^2(\Omega_T)$ as $m \rightarrow \infty$.

Proof. The idea of the proof is to use the Kolmogorov-Riesz theorem [9, Theorem 4.26] as in the continuous case; see [10, Section 4.4] or [22, Lemma 13]. The discrete case, however,

makes necessary some changes in the calculations. We need to show that

$$(48) \quad \lim_{(\xi, \tau) \rightarrow 0} \int_0^{T-\tau} \int_{\omega} \left((y_m z_m)(x + \xi, t + \tau) - (y_m z_m)(x, t) \right)^2 dx dt = 0$$

uniformly in m , where $\omega \subset \Omega$ satisfies $x + \xi \in \Omega$ for all $x \in \omega$. First, we separate the space and time translation:

$$\begin{aligned} & \int_0^{T-\tau} \int_{\omega} \left((y_m z_m)(x + \xi, t + \tau) - (y_m z_m)(x, t) \right)^2 dx dt \\ & \leq 2 \int_0^{T-\tau} \int_{\omega} \left((y_m z_m)(x + \xi, t + \tau) - (y_m z_m)(x, t + \tau) \right)^2 dx dt \\ & \quad + 2 \int_0^{T-\tau} \int_{\omega} \left((y_m z_m)(x, t + \tau) - (y_m z_m)(x, t) \right)^2 dx dt =: I_1 + I_2. \end{aligned}$$

For the estimate of I_1 , we apply a result for space translations of piecewise constant functions v with uniform bounds in the discrete $H^1(\Omega)$ norm, namely

$$\|v(\cdot + \xi) - v\|_{L^2(\Omega)}^2 \leq |\xi|(|\xi| + Ch(\mathcal{T}))\|v\|_{1,\mathcal{T}}^2$$

for appropriate ξ , where $C > 0$ only depends on Ω [15, Lemma 4]. This shows that

$$I_1 \leq C_1 |\xi|(|\xi| + Ch(\mathcal{T}_m))$$

converges to zero as $\xi \rightarrow 0$ uniformly in m .

For the second integral I_2 , we write

$$\begin{aligned} I_2 & \leq 4 \int_0^{T-\tau} \int_{\omega} z_m(x, t + \tau)^2 (y_m(x, t + \tau) - y_m(x, t))^2 dx dt \\ & \quad + 4 \int_0^{T-\tau} \int_{\omega} y_m(x, t)^2 (z_m(x, t + \tau) - z_m(x, t))^2 dx dt =: I_{21} + I_{22}. \end{aligned}$$

The L^∞ bounds on z_m give

$$I_{21} \leq C \int_0^{T-\tau} \int_{\omega} (y_m(x, t + \tau) - y_m(x, t))^2 dx dt.$$

By assumption, the sequence (y_m) is relatively compact in $L^2(\Omega_T)$. Therefore, we can apply the inverse of the Kolmogorov-Riesz theorem [9, Exercice 4.34] to conclude that I_{21} converges to zero as $\tau \rightarrow 0$ uniformly in m .

The analysis of I_{22} is more involved. We split the integral in several parts:

$$\begin{aligned} I_{22} & = \int_0^{T-\tau} \int_{\omega} y_m(x, t)^2 z_m(x, t) (z_m(x, t) - z_m(x, t + \tau)) dx dt \\ & \quad + \int_0^{T-\tau} \int_{\omega} y_m(x, t + \tau)^2 z_m(x, t + \tau) (z_m(x, t + \tau) - z_m(x, t)) dx dt \\ & \quad + \int_0^{T-\tau} \int_{\omega} (y_m(x, t)^2 - y_m(x, t + \tau)^2) z_m(x, t + \tau) (z_m(x, t + \tau) - z_m(x, t)) dx dt \end{aligned}$$

$$=: J_1 + J_2 + J_3.$$

The first two integrals J_1 and J_2 are treated similarly as in [8, Lemma 3.11]. Indeed, let $\lceil s \rceil$ denote the smallest integer larger or equal to s . Defining $n_m(t) := \lceil t/\Delta t_m \rceil$, we can formulate

$$z_m(x, t + \tau) - z_m(x, t) = \sum_{k=n_m(t)+1}^{n_m(t+\tau)} (z_{m,K}^k - z_{m,K}^{k-1})$$

for $x \in K$, $0 \leq t \leq T - \tau$. With this formulation, we can bound J_1 , using the duality of $\|\cdot\|_{1,\mathcal{T}_m}$ and $\|\cdot\|_{-1,\mathcal{T}_m}$:

$$\begin{aligned} J_1 &\leq \int_0^{T-\tau} \left(\sum_{K \in \mathcal{T}_m} m(K) (y_m^{n_m(t)})^2 z_{m,K}^{n_m(t)} \sum_{k=n_m(t)+1}^{n_m(t+\tau)} (z_{m,K}^{k-1} - z_{m,K}^k) \right) dt \\ &\leq \int_0^{T-\tau} \left(\sum_{k=n_m(t)+1}^{n_m(t+\tau)} \|(y_m^{n_m(t)})^2 z_m^{n_m(t)}\|_{1,\mathcal{T}_m} \|z_m^k - z_m^{k-1}\|_{-1,\mathcal{T}_m} \right) dt \\ &\leq \frac{1}{2} \int_0^{T-\tau} \sum_{k=n_m(t)+1}^{n_m(t+\tau)} \Delta t_m \|(y_m^{n_m(t)})^2 z_m^{n_m(t)}\|_{1,\mathcal{T}_m}^2 dt \\ &\quad + \frac{1}{2} \int_0^{T-\tau} \sum_{k=n_m(t)+1}^{n_m(t+\tau)} \frac{1}{\Delta t_m} \|z_m^k - z_m^{k-1}\|_{-1,\mathcal{T}_m}^2 dt \\ &\leq \frac{\tau}{2} \sum_{k=1}^{N_m} \Delta t_m \|(y_m^k)^2 z_m^k\|_{1,\mathcal{T}_m}^2 + \frac{\tau}{2} \sum_{k=1}^{N_m} \frac{1}{\Delta t_m} \|z_m^k - z_m^{k-1}\|_{-1,\mathcal{T}_m}^2, \end{aligned}$$

where the last inequality follows from [5, Lemmas 4.1 and 4.2]. Let us remark that, for all $\sigma = K|L \in \mathcal{E}_{\text{int}}$, we can rewrite

$$(y_K)^2 z_K - (y_L)^2 z_L = \frac{y_K + y_L}{2} (y_K z_K - y_L z_L) + \frac{y_K z_K + y_L z_L}{2} (y_K - y_L).$$

Then,

$$\|(y_m^k)^2 z_m^k\|_{1,\mathcal{T}_m}^2 \leq m(\Omega) \|(y_m^k)^2 z_m^k\|_{L^\infty(\Omega)}^2 + \|y_m^k\|_{L^\infty(\Omega)} \|y_m^k z_m^k\|_{1,\mathcal{T}_m} + \|y_m^k z_m^k\|_{L^\infty(\Omega)} \|y_m^k\|_{1,\mathcal{T}_m}.$$

Hence, $J_1 \leq C\tau$ for some $C > 0$. An analogous estimation leads to $J_2 \leq C\tau$. It remains to estimate the integral J_3 . For this, we use, similar to the treatment of I_{21} , the L^∞ bounds on y_m and z_m :

$$J_3 \leq C \int_0^{T-\tau} \int_\omega |y_m(x, t + \tau) - y_m(x, t)| dx dt.$$

This expression converges to zero uniformly in m because of the relative compactness of (y_m) in $L^2(\Omega_T)$.

We deduce from the previous computations that (48) holds true. Therefore, the product $(y_m z_m)$ converges strongly in $L^2(\Omega_T)$, up to some subsequence, and in view of the convergences $y_m \rightarrow y$ strongly in $L^2(\Omega_T)$ and $z_m \rightharpoonup^* z$ weakly* in $L^\infty(0, T; L^\infty(\Omega))$, the limit of $(y_m z_m)$ equals yz , which finishes the proof. \square

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